DonorsChoose

DonorsChoose.org receives hundreds of thousands of project proposals each year for classroom projects in need of funding. Right now, a large number of volunteers is needed to manually screen each submission before it's approved to be posted on the DonorsChoose.org website.

Next year, DonorsChoose.org expects to receive close to 500,000 project proposals. As a result, there are three main problems they need to solve:

- How to scale current manual processes and resources to screen 500,000 projects so that they can be posted as quickly and as efficiently as possible
- · How to increase the consistency of project vetting across different volunteers to improve the experience for teachers
- How to focus volunteer time on the applications that need the most assistance

The goal of the competition is to predict whether or not a DonorsChoose.org project proposal submitted by a teacher will be approved, using the text of project descriptions as well as additional metadata about the project, teacher, and school. DonorsChoose.org can then use this information to identify projects most likely to need further review before approval.

About the DonorsChoose Data Set

The train.csv data set provided by DonorsChoose contains the following features:

Feature	Description
project_id	A unique identifier for the proposed project. Example: p036502
	Title of the project. Examples:
project_title	Art Will Make You Happy!
	• First Grade Fun
	Grade level of students for which the project is targeted. One of the following enumerated values:
project grade category	• Grades PreK-2
project_grade_category	• Grades 3-5
	• Grades 6-8
	• Grades 9-12
	One or more (comma-separated) subject categories for the project from the following enumerated list of values:
	Applied Learning
	• Care & Hunger
	• Health & Sports
	History & Civics
	• Literacy & Language
project_subject_categories	• Math & Science
	• Music & The Arts
	• Special Needs
	• Warmth
	Examples:
	• Music & The Arts
	• Literacy & Language, Math & Science
school_state	State where school is located (<u>Two-letter U.S. postal code</u>). Example
	One or more (comma-separated) subject subcategories for the project
project_subject_subcategories	Examples:
	• Literacy

Feature	• Literature & Writing, Social Sciences Description		
project_resource_summary	An explanation of the resources needed for the project. Example: • My students need hands on literacy materials to manage sensory needs!		
project_essay_1	First application essay [*]		
project_essay_2	Second application essay*		
project_essay_3	Third application essay*		
project_essay_4	Fourth application essay*		
project_submitted_datetime	Datetime when project application was submitted. Example: 2016–04–28 12:43:56.245		
teacher_id	A unique identifier for the teacher of the proposed project. Example: bdf8baa8fedef6bfeec7ae4ff1c15c56		
teacher_prefix	Teacher's title. One of the following enumerated values: • nan • Dr. • Mr. • Mrs. • Ms. • Teacher.		
teacher_number_of_previously_posted_projects	Number of project applications previously submitted by the same teacher. Example: 2		

^{*} See the section **Notes on the Essay Data** for more details about these features.

Additionally, the resources.csv data set provides more data about the resources required for each project. Each line in this file represents a resource required by a project:

Feature	Description			
id	A project_id value from the train.csv file. Example: p036502			
description	Desciption of the resource. Example: Tenor Saxophone Reeds, Box of 25			
quantity	Quantity of the resource required. Example: 3			
price	Price of the resource required. Example: 9.95			

Note: Many projects require multiple resources. The id value corresponds to a project_id in train.csv, so you use it as a key to retrieve all resources needed for a project:

The data set contains the following label (the value you will attempt to predict):

Label	Description	
project is approved	A binary flag indicating whether DonorsChoose approved the project. A value of 0 indicates the project	
project_is_approved	was not approved, and a value of 1 indicates the project was approved.	

Notes on the Essay Data

Prior to May 17, 2016, the prompts for the essays were as follows:

- __project_essay_1:__ "Introduce us to your classroom"
- __project_essay_2:__ "Tell us more about your students"
- __project_essay_3:__ "Describe how your students will use the materials you're requesting"
- __project_essay_3:__ "Close by sharing why your project will make a difference"

Starting on May 17, 2016, the number of essays was reduced from 4 to 2, and the prompts for the first 2 essays were changed to the following:

• __project_essay_1:__ "Describe your students: What makes your students special? Specific details about their background, your neighborhood, and your school are all helpful."

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 __project_essay_2:__ "About your project: How will these materials make a difference in your students' learning and improve their school lives?"

For all projects with project_submitted_datetime of 2016-05-17 and later, the values of project_essay_3 and project_essay_4 will be NaN.

1.1 Reading Data

```
In [49]:
```

```
%matplotlib inline
import warnings
warnings.filterwarnings("ignore")
import sqlite3
import pandas as pd
import numpy as np
import nltk
import string
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.feature_extraction.text import TfidfTransformer
from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.feature_extraction.text import CountVectorizer
from sklearn.metrics import confusion matrix
from sklearn import metrics
from sklearn.metrics import roc curve, auc
from nltk.stem.porter import PorterStemmer
import re
 Tutorial about Python regular expressions: https://pymotw.com/2/re/
import string
from nltk.corpus import stopwords
from nltk.stem import PorterStemmer
from nltk.stem.wordnet import WordNetLemmatizer
from gensim.models import Word2Vec
from gensim.models import KeyedVectors
import pickle
from tqdm import tqdm
import os
from plotly import plotly
import plotly.offline as offline
import plotly.graph_objs as go
offline.init notebook mode()
from collections import Counter
```

```
In [50]:
```

```
project_data = pd.read_csv('train_data.csv')
resource_data = pd.read_csv('resources.csv')
```

In [51]:

```
print("Number of data points in train data", project_data.shape)
print('-'*50)
print("The attributes of data :", project_data.columns.values)
```

In [52]:

```
print("Number of data points in train data", resource_data.shape)
print(resource_data.columns.values)
resource_data.head(2)

Number of data points in train data (1541272, 4)
['id' 'description' 'quantity' 'price']
```

Out[52]:

	id	quantity	price	
0	p233245	LC652 - Lakeshore Double-Space Mobile Drying Rack	1	149.00
1	p069063	Bouncy Bands for Desks (Blue support pipes)	3	14.95

1.2 preprocessing of project subject categories

In [53]:

```
catogories = list(project data['project subject categories'].values)
# remove special characters from list of strings python:
https://stackoverflow.com/a/47301924/4084039
# https://www.geeksforgeeks.org/removing-stop-words-nltk-python/
# https://stackoverflow.com/questions/23669024/how-to-strip-a-specific-word-from-a-string
# https://stackoverflow.com/questions/8270092/remove-all-whitespace-in-a-string-in-python
cat_list = []
for i in catogories:
    temp = ""
    # consider we have text like this "Math & Science, Warmth, Care & Hunger"
    for j in i.split(','): # it will split it in three parts ["Math & Science", "Warmth", "Care & E
       if 'The' in j.split(): # this will split each of the catogory based on space "Math & Science"
e"=> "Math","&", "Science"
            j=j.replace('The','') # if we have the words "The" we are going to replace it with ''(i
.e removing 'The')
       j = j.replace(' ','') # we are placeing all the ' '(space) with ''(empty) ex:"Math &
Science"=>"Math&Science"
       temp+=j.strip()+" " #" abc ".strip() will return "abc", remove the trailing spaces
        temp = temp.replace('&','_') # we are replacing the & value into
    cat_list.append(temp.strip())
project data['clean categories'] = cat list
project_data.drop(['project_subject_categories'], axis=1, inplace=True)
from collections import Counter
my counter = Counter()
for word in project data['clean categories'].values:
   my_counter.update(word.split())
cat dict = dict(my counter)
sorted cat dict = dict(sorted(cat dict.items(), key=lambda kv: kv[1]))
4
```

1.3 preprocessing of project_subject_subcategories

In [54]:

```
# consider we have text like this "Math & Science, Warmth, Care & Hunger"
    for j in i.split(','): # it will split it in three parts ["Math & Science", "Warmth", "Care & E
unger"]
       if 'The' in j.split(): # this will split each of the catogory based on space "Math & Science"
e"=> "Math","&", "Science"
            j=j.replace('The','') # if we have the words "The" we are going to replace it with ''(i
.e removing 'The')
       j = j.replace(' ','') # we are placeing all the ' '(space) with ''(empty) ex:"Math &
Science"=>"Math&Science"
       temp +=j.strip()+" "#" abc ".strip() will return "abc", remove the trailing spaces
        temp = temp.replace('&',' ')
    sub cat list.append(temp.strip())
project_data['clean_subcategories'] = sub_cat_list
project data.drop(['project subject subcategories'], axis=1, inplace=True)
# count of all the words in corpus python: https://stackoverflow.com/a/22898595/4084039
my counter = Counter()
for word in project_data['clean_subcategories'].values:
   my_counter.update(word.split())
sub cat dict = dict(my counter)
sorted sub cat dict = dict(sorted(sub cat dict.items(), key=lambda kv: kv[1]))
```

1.4 preprocessing of project_grade_category

```
In [55]:
```

```
prj grade cat = list(project data['project grade category'].values)
# remove special characters from list of strings python:
https://stackoverflow.com/a/47301924/4084039
# https://www.geeksforgeeks.org/removing-stop-words-nltk-python/
# https://stackoverflow.com/questions/23669024/how-to-strip-a-specific-word-from-a-string
# https://stackoverflow.com/questions/8270092/remove-all-whitespace-in-a-string-in-python
prj_grade_cat_list = []
for i in prj_grade_cat:
    for j in i.split(' '): # it will split by space
       j=j.replace('Grades','') # if we have the words "Grades" we are going to replace it with ''
(i.e removing 'Grades')
   prj grade cat list.append(j.strip())
project data['clean grade'] = prj grade cat list
project_data.drop(['project_grade_category'], axis=1, inplace=True)
# count of all the words in corpus python: https://stackoverflow.com/a/22898595/4084039
my counter = Counter()
for word in project data['clean grade'].values:
   my_counter.update(word.split())
prj_grade_cat_dict = dict(my_counter)
sorted_prj_grade_cat_dict = dict(sorted(prj_grade_cat_dict.items(), key=lambda kv: kv[1]))
project data['clean grade'].values
Out [551:
array(['PreK-2', '6-8', '6-8', ..., 'PreK-2', '3-5', '6-8'], dtype=object)
```

1.5 preprocessing of teacher_prefix

```
In [56]:
```

```
#tea_pfx_cat = list(project_data['teacher_prefix'].values)
tea_pfx_cat = list(project_data['teacher_prefix'].astype(str).values)
# remove special characters from list of strings python:
https://stackoverflow.com/a/47301924/4084039

# https://www.geeksforgeeks.org/removing-stop-words-nltk-python/
# https://stackoverflow.com/questions/23669024/how-to-strip-a-specific-word-from-a-string
```

```
# https://stackoverflow.com/questions/8270092/remove-all-whitespace-in-a-string-in-python
##https://stackoverflow.com/questions/52736900/how-to-solve-the-attribute-error-float-object-has-n
o-attribute-split-in-pyth
#vectorizer.fit(project_data['teacher_prefix'].astype(str).values)
tea_pfx_cat_list = []
for i in tea_pfx_cat:
    #for j in i.split(' '): # it will split by space
    \#j=j.replace('.','') \# if we have the words "Grades" we are going to replace it with ''(i.e re
moving 'Grades')
   i=i.replace('.','') # if we have the words "Grades" we are going to replace it with ''(i.e remc
ving 'Grades')
   i=i.replace('nan','') # if we have the words "Grades" we are going to replace it with ''(i.e re
moving 'Grades')
    tea pfx cat list.append(i.strip())
project data['clean tea pfx'] = tea pfx cat list
project data.drop(['teacher prefix'], axis=1, inplace=True)
# count of all the words in corpus python: https://stackoverflow.com/a/22898595/4084039
my counter = Counter()
for word in project data['clean tea pfx'].values:
   my_counter.update(word.split())
tea pfx cat dict = dict(my counter)
sorted_tea_pfx_cat_dict = dict(sorted(tea_pfx_cat_dict.items(), key=lambda kv: kv[1]))
project data['clean tea pfx'].values
4
Out [56]:
array(['Mrs', 'Mr', 'Ms', ..., 'Mrs', 'Mrs', 'Ms'], dtype=object)
```

1.6 Text preprocessing

In [57]:

```
In [58]:
```

```
project_data.head(2)
```

Out[58]:

	Unnamed:	id	teacher_id	school_state	project_submitted_datetime	project_title	projec
0	160221	p253737	c90749f5d961ff158d4b4d1e7dc665fc	IN	2016-12-05 13:43:57	Educational Support for English Learners at Home	My stu Englisl that ar
1	140945	p258326	897464ce9ddc600bced1151f324dd63a	FL	2016-10-25 09:22:10	Wanted: Projector for Hungry Learners	Our stuarrive school lea

Using Pretrained Models: TFIDF weighted W2V

In [59]:

```
# printing some random reviews
print(project_data['essay'].values[0])
print("="*50)
print(project_data['essay'].values[150])
print(project_data['essay'].values[1000])
print("="*50)
print(project_data['essay'].values[20000])
print("="*50)
print(project_data['essay'].values[99999])
print(project_data['essay'].values[99999])
print("="*50)
```

My students are English learners that are working on English as their second or third languages. W e are a melting pot of refugees, immigrants, and native-born Americans bringing the gift of langua ge to our school. \r We have over 24 languages represented in our English Learner program wi th students at every level of mastery. We also have over 40 countries represented with the families within our school. Each student brings a wealth of knowledge and experiences to us that open our eyes to new cultures, beliefs, and respect.\"The limits of your language are the limits o f your world.\"-Ludwig Wittgenstein Our English learner's have a strong support system at home th at begs for more resources. Many times our parents are learning to read and speak English along s ide of their children. Sometimes this creates barriers for parents to be able to help their child learn phonetics, letter recognition, and other reading skills.\r\n\r\nBy providing these dvd's and players, students are able to continue their mastery of the English language even if no one at hom e is able to assist. All families with students within the Level 1 proficiency status, will be a offered to be a part of this program. These educational videos will be specially chosen by the En glish Learner Teacher and will be sent home regularly to watch. The videos are to help the child develop early reading skills.\r\n\rangle parents that do not have access to a dvd player will have the opportunity to check out a dvd player to use for the year. The plan is to use these videos and ed ucational dvd's for the years to come for other EL students.\r\nnannan

The 51 fifth grade students that will cycle through my classroom this year all love learning, at 1 east most of the time. At our school, 97.3% of the students receive free or reduced price lunch. O f the 560 students, 97.3% are minority students. \r\nThe school has a vibrant community that loves to get together and celebrate. Around Halloween there is a whole school parade to show off the bea utiful costumes that students wear. On Cinco de Mayo we put on a big festival with crafts made by the students, dances, and games. At the end of the year the school hosts a carnival to celebrate t he hard work put in during the school year, with a dunk tank being the most popular activity.My st udents will use these five brightly colored Hokki stools in place of regular, stationary, 4-legged chairs. As I will only have a total of ten in the classroom and not enough for each student to hav e an individual one, they will be used in a variety of ways. During independent reading time they will be used as special chairs students will each use on occasion. I will utilize them in place of chairs at my small group tables during math and reading times. The rest of the day they will be us ed by the students who need the highest amount of movement in their life in order to stay focused on $school.\rdot n\rdot n\rdo$ Stools. They can't get their fill of the 5 stools we already have. When the students are sitting i n group with me on the Hokki Stools, they are always moving, but at the same time doing their work. Anytime the students get to pick where they can sit, the Hokki Stools are the first to be ta ken. There are always students who head over to the kidney table to get one of the stools who are disappointed as there are not enough of them. \r\n\we ask a lot of students to sit for 7 hours a day. The Hokki stools will be a compromise that allow my students to do desk work and move at th e same time. These stools will help students to meet their 60 minutes a day of movement by allowing them to activate their core muscles for balance while they sit. For many of my students, these chairs will take away the barrier that exists in schools for a child who can't sit still.nannan

_____ How do you remember your days of school? Was it in a sterile environment with plain walls, rows of desks, and a teacher in front of the room? A typical day in our room is nothing like that. I work hard to create a warm inviting themed room for my students look forward to coming to each day.\r\n \r nMy class is made up of 28 wonderfully unique boys and girls of mixed races in Arkansas.\r\nThey attend a Title I school, which means there is a high enough percentage of free a nd reduced-price lunch to qualify. Our school is an \"open classroom\" concept, which is very uniq ue as there are no walls separating the classrooms. These 9 and 10 year-old students are very eage r learners; they are like sponges, absorbing all the information and experiences and keep on wanti ng more.With these resources such as the comfy red throw pillows and the whimsical nautical hangin q decor and the blue fish nets, I will be able to help create the mood in our classroom setting to be one of a themed nautical environment. Creating a classroom environment is very important in the success in each and every child's education. The nautical photo props will be used with each child as they step foot into our classroom for the first time on Meet the Teacher evening. I'll take pic tures of each child with them, have them developed, and then hung in our classroom ready for their first day of 4th grade. This kind gesture will set the tone before even the first day of school! The nautical thank you cards will be used throughout the year by the students as they create thank

you cards to their team groups.\r\n\r\nyour generous donations will help me to help make our classroom a fun, inviting, learning environment from day one.\r\n\r\nIt costs lost of money out of my own pocket on resources to get our classroom ready. Please consider helping with this project to make our new school year a very successful one. Thank you!nannan

My kindergarten students have varied disabilities ranging from speech and language delays, cognitive delays, gross/fine motor delays, to autism. They are eager beavers and always strive to work their hardest working past their limitations. \r\n\r\nThe materials we have are the ones I seek out for my students. I teach in a Title I school where most of the students receive free or reduced price lunch. Despite their disabilities and limitations, my students love coming to school and come eager to learn and explore. Have you ever felt like you had ants in your pants and you needed to grove and move as you were in a meeting? This is how my kids feel all the time. The want to be able to move as they learn or so they say. Wobble chairs are the answer and I love then because they develop their core, which enhances gross motor and in Turn fine motor skills. \r\nThey also want to learn through games, my kids don't want to sit and do worksheets. They want to learn to count by jumping and playing. Physical engagement is the key to our success. The number toss and color and shape mats can make that happen. My students will forget they are doing work and just have the fun a 6 year old deserves.nannan

The mediocre teacher tells. The good teacher explains. The superior teacher demonstrates. The grea t teacher inspires. -William A. Ward\r\n\r\nMy school has 803 students which is makeup is 97.6% Af rican-American, making up the largest segment of the student body. A typical school in Dallas is m ade up of 23.2% African-American students. Most of the students are on free or reduced lunch. We a ren't receiving doctors, lawyers, or engineers children from rich backgrounds or neighborhoods. As an educator I am inspiring minds of young children and we focus not only on academics but one smar t, effective, efficient, and disciplined students with good character. In our classroom we can util ize the Bluetooth for swift transitions during class. I use a speaker which doesn't amplify the so und enough to receive the message. Due to the volume of my speaker my students can't hear videos or books clearly and it isn't making the lessons as meaningful. But with the bluetooth speaker my students will be able to hear and I can stop, pause and replay it at any time.\r\nThe cart will all ow me to have more room for storage of things that are needed for the day and has an extra part to it I can use. The table top chart has all of the letter, words and pictures for students to learn about different letters and it is more accessible.nannan

In [60]:

```
# https://stackoverflow.com/a/47091490/4084039
import re
def decontracted(phrase):
   # specific
   phrase = re.sub(r"won't", "will not", phrase)
   phrase = re.sub(r"can\'t", "can not", phrase)
    # general
   phrase = re.sub(r"n\'t", " not", phrase)
   phrase = re.sub(r"\'re", " are", phrase)
   phrase = re.sub(r"\'s", " is", phrase)
   phrase = re.sub(r"\'d", " would", phrase)
   phrase = re.sub(r"\'ll", " will", phrase)
   phrase = re.sub(r"\'t", " not", phrase)
   phrase = re.sub(r"\'ve", " have", phrase)
   phrase = re.sub(r"\'m", " am", phrase)
   return phrase
```

In [61]:

```
sent = decontracted(project_data['essay'].values[20000])
print(sent)
print("="*50)
```

My kindergarten students have varied disabilities ranging from speech and language delays, cognitive delays, gross/fine motor delays, to autism. They are eager beavers and always strive to work their hardest working past their limitations. \r\n\r\nThe materials we have are the ones I seek out for my students. I teach in a Title I school where most of the students receive free or reduced price lunch. Despite their disabilities and limitations, my students love coming to school and come eager to learn and explore. Have you ever felt like you had ants in your pants and you needed to grove and move as you were in a meeting? This is how my kids feel all the time. The want to be able to move as they learn or so they say. Wobble chairs are the answer and I love then because they develop their core, which enhances gross motor and in Turn fine motor skills. \r\nThey also want to learn through games, my kids do not want to sit and do worksheets. They want to learn to count by jumping and playing. Physical engagement is the key to our success. The number toss and color and shape mats can make that happen. My students will forget they are doing work and just have the funda 6 year old deserves.nannan

In [62]:

```
# \r \n \t remove from string python: http://texthandler.com/info/remove-line-breaks-python/
sent = sent.replace('\\r', ' ')
sent = sent.replace('\\"', ' ')
sent = sent.replace('\\n', ' ')
print(sent)
```

My kindergarten students have varied disabilities ranging from speech and language delays, cognitive delays, gross/fine motor delays, to autism. They are eager beavers and always strive to work their hardest working past their limitations. The materials we have are the ones I seek out for my students. I teach in a Title I school where most of the students receive free or reduced price lunch. Despite their disabilities and limitations, my students love coming to school and come eager to learn and explore. Have you ever felt like you had ants in your pants and you needed to groove and move as you were in a meeting? This is how my kids feel all the time. The want to be able to move as they learn or so they say. Wobble chairs are the answer and I love then because they develop their core, which enhances gross motor and in Turn fine motor skills. They also want to learn through games, my kids do not want to sit and do worksheets. They want to learn to count by jumping and playing. Physical engagement is the key to our success. The number toss and color and shape mats can make that happen. My students will forget they are doing work and just have the fun a 6 year old deserves.nannan

In [63]:

```
#remove spacial character: https://stackoverflow.com/a/5843547/4084039
sent = re.sub('[^A-Za-z0-9]+', ' ', sent)
print(sent)
```

My kindergarten students have varied disabilities ranging from speech and language delays cognitive delays gross fine motor delays to autism They are eager beavers and always strive to work their hardest working past their limitations. The materials we have are the ones I seek out for my students I teach in a Title I school where most of the students receive free or reduced price lunch. Despite their disabilities and limitations my students love coming to school and come eager to learn and explore Have you ever felt like you had ants in your pants and you needed to groove and move as you were in a meeting This is how my kids feel all the time. The want to be able to move as the ey learn or so they say Wobble chairs are the answer and I love then because they develop their come which enhances gross motor and in Turn fine motor skills. They also want to learn through games my kids do not want to sit and do worksheets. They want to learn to count by jumping and playing Pheysical engagement is the key to our success. The number toss and color and shape mats can make that happen My students will forget they are doing work and just have the fun a 6 year old deserves nan nan

In [64]:

```
# https://gist.github.com/sebleier/554280
# we are removing the words from the stop words list: 'no', 'nor', 'not'
stopwords= ['i', 'me', 'my', 'myself', 'we', 'our', 'ours', 'ourselves', 'you', "you're", "you've",
                            "you'll", "you'd", 'your', 'yours', 'yourself', 'yourselves', 'he', 'him', 'his',
'himself', \
                            'she', "she's", 'her', 'hers', 'herself', 'it', "it's", 'its', 'itself', 'they', 'them',
 'their'.\
                            'theirs', 'themselves', 'what', 'which', 'who', 'whom', 'this', 'that', "that'll",
'these', 'those', \
                             'am', 'is', 'are', 'was', 'were', 'be', 'been', 'being', 'have', 'has', 'had', 'having',
'do', 'does', \
                             'did', 'doing', 'a', 'an', 'the', 'and', 'but', 'if', 'or', 'because', 'as', 'until', '
while', 'of', \
                             'at', 'by', 'for', 'with', 'about', 'against', 'between', 'into', 'through', 'during',
'before', 'after',\
                            'above', 'below', 'to', 'from', 'up', 'down', 'in', 'out', 'on', 'off', 'over', 'under'
, 'again', 'further',\
                            'then', 'once', 'here', 'there', 'when', 'why', 'how', 'all', 'any', 'both', '\epsilon
ach', 'few', 'more',\
                            'most', 'other', 'some', 'such', 'only', 'own', 'same', 'so', 'than', 'too', 'very', \
                            's', 't', 'can', 'will', 'just', 'don', "don't", 'should', "should've", 'now', 'd', 'll'
, 'm', 'o', 're', \
                            've', 'y', 'ain', 'aren', "aren't", 'couldn', "couldn't", 'didn', "didn't", 'doesn', "doesn', "doesn',
esn't", 'hadn',\
                            "hadn't", 'hasn', "hasn't", 'haven', "haven't", 'isn', "isn't", 'ma', 'mightn',
"mightn't". 'mustn'.\
```

```
"mustn't", 'needn', "needn't", 'shan', "shan't", 'shouldn', "shouldn't", 'wasn',
"wasn't", 'weren', "weren't", \
           'won', "won't", 'wouldn', "wouldn't"]
In [65]:
# Combining all the above stundents
from tqdm import tqdm
preprocessed essays = []
# tqdm is for printing the status bar
for sentance in tqdm (project data['essay'].values):
   sent = decontracted(sentance)
   sent = sent.replace('\\r', ' ')
   sent = sent.replace('\\"', ' ')
   sent = sent.replace('\\n', ' ')
   sent = re.sub('[^A-Za-z0-9]+', '', sent)
   # https://gist.github.com/sebleier/554280
    sent = ' '.join(e for e in sent.split() if e not in stopwords)
   preprocessed_essays.append(sent.lower().strip())
                                                                            | 109248/109248
[01:01<00:00, 1776.34it/s]
```

In [66]:

```
# after preprocesing
preprocessed_essays[20000]
```

Out[66]:

'my kindergarten students varied disabilities ranging speech language delays cognitive delays gros s fine motor delays autism they eager beavers always strive work hardest working past limitations the materials ones i seek students i teach title i school students receive free reduced price lunc h despite disabilities limitations students love coming school come eager learn explore have ever felt like ants pants needed groove move meeting this kids feel time the want able move learn say w obble chairs answer i love develop core enhances gross motor turn fine motor skills they also want learn games kids not want sit worksheets they want learn count jumping playing physical engagement key success the number toss color shape mats make happen my students forget work fun 6 year old de serves nannan'

1.7 Preprocessing of `project_title`

In [67]:

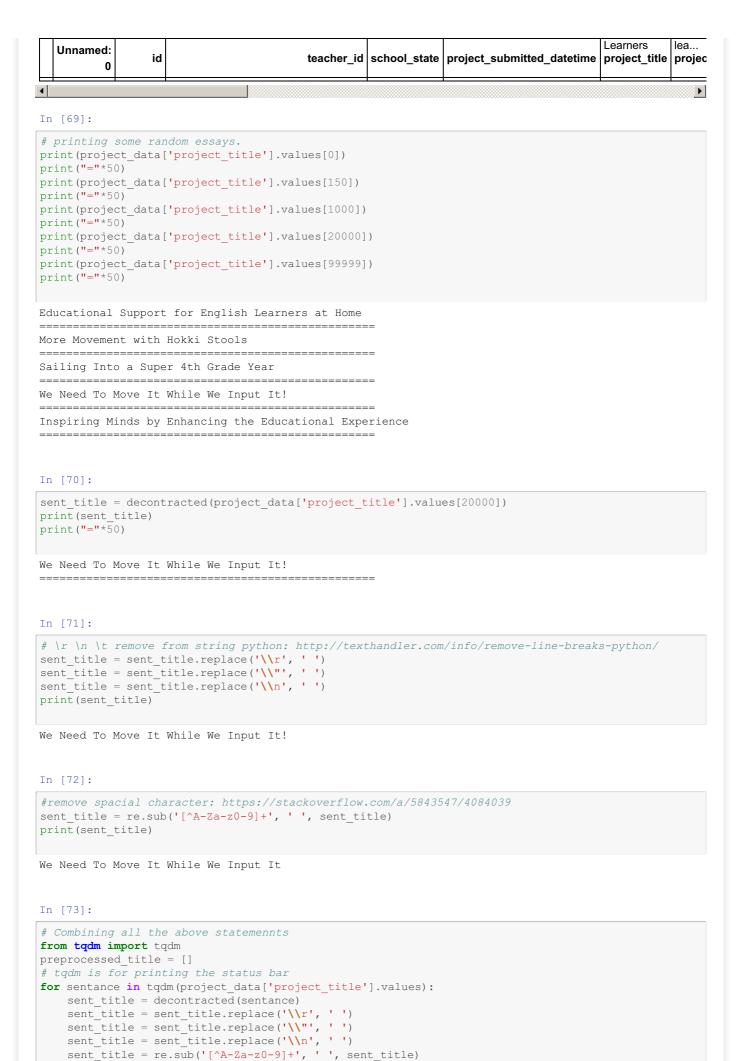
```
# similarly you can preprocess the titles also
```

In [68]:

```
project_data.head(2)
```

Out[68]:

	Unnamed: 0	id	teacher_id	school_state	project_submitted_datetime	project_title	projec
0	160221	p253737	c90749f5d961ff158d4b4d1e7dc665fc	IN	2016-12-05 13:43:57	English	My stu Englisi that ar
1	140945	p258326	897464ce9ddc600bced1151f324dd63a	FL	2016-10-25 09:22:10	Projector for	Our stu



https://gist.github.com/sebleier/554280

sent title = ' '.join(e for e in sent title.split() if e not in stopwords)

```
preprocessed title.append(sent title.lower().strip())
                                                                            109248/109248
100%।
[00:02<00:00, 37550.40it/s]
In [74]:
# after preprocesing
preprocessed title[10]
Out[74]:
'reading changes lives'
In [75]:
# Combining all the above statemennts
from tqdm import tqdm
preprocessed prj sum = []
# tqdm is for printing the status bar
for sentance in tqdm(project_data['project_resource_summary'].values):
   sent title = decontracted(sentance)
    sent_title = sent_title.replace('\\r', ' ')
    sent_title = sent_title.replace('\\"', ' ')
    sent_title = sent_title.replace('\\n', ' ')
    sent_title = re.sub('[^A-Za-z0-9]+', ' ', sent_title)
    # https://gist.github.com/sebleier/554280
    sent_title = ' '.join(e for e in sent_title.split() if e not in stopwords)
    preprocessed_prj_sum.append(sent_title.lower().strip())
                                                                          | 109248/109248
[00:06<00:00, 15947.81it/s]
```

1.8 Numeric feature for Text

1.8.1 Numerric feature for essay

```
In [76]:
```

141

In [77]:

```
# Suggestion 5.you can try improving the score using feature engineering hacks.Try including lengt
h,summary
# and observe the results and re-submit the assignment.

# https://stackoverflow.com/questions/18827198/python-count-number-of-words-in-a-list-strings
preprocessed_essays_len = []
for item in tqdm(preprocessed_essays):
    preprocessed_essays_len.append(len(item))

print(preprocessed_essays_len[101])
```

```
100%| 100%| 1009248/109248 [00:00<00:00, 1343921.15it/s]
```

1.8.2 Numerric feature for title

```
In [78]:
```

```
# Suggestion 5.you can try improving the score using feature engineering hacks.Try including lengt
h,summary
# and observe the results and re-submit the assignment.

# https://stackoverflow.com/questions/18827198/python-count-number-of-words-in-a-list-strings
preprocessed_title_wc = []
for item in tqdm(preprocessed_title):
    preprocessed_title_wc.append(len(item.split()))

print(preprocessed_title_wc[101])

100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 10
```

In [79]:

```
# Suggestion 5.you can try improving the score using feature engineering hacks. Try including lengt h, summary
# and observe the results and re-submit the assignment.

# https://stackoverflow.com/questions/18827198/python-count-number-of-words-in-a-list-strings
preprocessed_title_len = []

for item in tqdm(preprocessed_title):
    #print(preprocessed_title)
    preprocessed_title_len.append(len(item))
    #print(len(preprocessed_title))

print(preprocessed_title_len[101])

100%| 100<00:00, 1711421.16it/s]
```

18

1.8.2 Numerric feature for project_summary_resource

```
In [80]:
```

```
# Suggestion 5.you can try improving the score using feature engineering hacks.Try including lengt
h,summary
# and observe the results and re-submit the assignment.

# https://stackoverflow.com/questions/18827198/python-count-number-of-words-in-a-list-strings
preprocessed_prj_sum_wc = []
for item in tqdm(preprocessed_prj_sum):
    preprocessed_prj_sum_wc.append(len(item.split()))

print(preprocessed_prj_sum_wc[100])

100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%| 100%|
```

```
Out[82]:
Index(['Unnamed: 0', 'id', 'teacher_id', 'school_state',
       'project_submitted_datetime', 'project_title', 'project_essay_1',
       'project_essay_2', 'project_essay_3', 'project_essay_4',
       'project_resource_summary',
       'teacher_number_of_previously_posted_projects', 'project_is_approved',
       'clean_categories', 'clean_subcategories', 'clean_grade',
       'clean_tea_pfx', 'essay'],
      dtype='object')
we are going to consider
      - school_state : categorical data
      - clean categories : categorical data
      - clean subcategories : categorical data
      - project_grade_category : categorical data
      - teacher_prefix : categorical data
      - project title : text data
      - text : text data
      - project resource summary: text data (optinal)
      - quantity : numerical (optinal)
      - teacher_number_of_previously_posted_projects : numerical
      - price : numerical
```

Using Pretrained Models: Avg W2V

```
In [83]:
```

```
# Reading glove vectors in python: https://stackoverflow.com/a/38230349/4084039

def loadGloveModel(gloveFile):
    print ("Loading Glove Model")
    f = open(gloveFile,'r', encoding="utf8")
    model = {}
    for line in tqdm(f):
        splitLine = line.split()
        word = splitLine[0]
        embedding = np.array([float(val) for val in splitLine[1:]])
        model[word] = embedding
    print ("Done.",len(model)," words loaded!")

return model
```

```
model = loadGloveModel('glove.42B.300d.txt')
 # -----
Output:
Loading Glove Model
 1917495it [06:32, 4879.69it/s]
Done. 1917495 words loaded!
 # -----
words = []
 for i in preproced texts:
        words.extend(i.split(' '))
for i in preproced titles:
        words.extend(i.split(' '))
print("all the words in the coupus", len(words))
words = set(words)
print("the unique words in the coupus", len(words))
inter_words = set(model.keys()).intersection(words)
print("The number of words that are present in both glove vectors and our coupus", \
              len(inter words),"(",np.round(len(inter words)/len(words)*100,3),"%)")
words courpus = {}
words_glove = set(model.keys())
for i in words:
        if i in words glove:
                 words courpus[i] = model[i]
print("word 2 vec length", len(words courpus))
 # stronging variables into pickle files python: http://www.jessicayung.com/how-to-use-pickle-to-sa
ve-and-load-variables-in-python/
 import pickle
with open('glove_vectors', 'wb') as f:
      pickle.dump(words courpus, f)
 . . .
Out[83]:
\label{loadGloveModel} \begin{tabular}{ll} \
encoding="utf8")\n model = {}\n for line in tqdm(f):\n
                                                                                                                                                   splitLine = line.split()\n
                                                       embedding = np.array([float(val) for val in splitLine[1:]])\n
word = splitLine[0]\n
odel[word] = embedding\n print ("Done.",len(model)," words loaded!")\n return model\nmodel =
\label{loadGloveModel('glove.42B.300d.txt'') \n\fi} = = = = = = ---- \\ \n\c G \n\c G
love Model\n1917495it [06:32, 4879.69it/s]\nDone. 1917495 words loaded!\n\n#
-----\n\nwords = []\nfor i in preproced_texts:\n
                                                                                                                                                                    words.extend(i.split(\'
coupus", len(words)) \nwords = set(words) \nprint("the unique words in the coupus",
len(words)) \n\ninter words = set(model.keys()).intersection(words) \nprint("The number of words tha
t are present in both glove vectors and our coupus",
                                                                                                                              len(inter words),"
(",np.round(len(inter_words)/len(words)*100,3),"%)")\n\nwords_courpus = {}\nwords_glove =
print("word 2 vec length", len(words courpus)) \n\n# stronging variables into pickle files python
: http://www.jessicayung.com/how-to-use-pickle-to-save-and-load-variables-in-python/\n\nimport pic
4
                                                                                                                                                                                                                    Þ
In [841:
 # stronging variables into pickle files python: http://www.jessicayung.com/how-to-use-pickle-to-sa
ve-and-load-variables-in-python/
 # make sure you have the glove vectors file
with open('glove vectors', 'rb') as f:
        model = pickle.load(f)
        glove_words = set(model.keys())
```

тегити шоает

In [85]:

```
## https://monkeylearn.com/sentiment-analysis/
## http://t-redactyl.io/blog/2017/04/using-vader-to-handle-sentiment-analysis-with-social-media-te
xt.html
#import nltk
#from nltk.sentiment.vader import SentimentIntensityAnalyzer
#import nltk
#nltk.download('vader lexicon')
#sid = SentimentIntensityAnalyzer()
#for sentiment = 'a person is a person no matter how small dr seuss i teach the smallest students
with the biggest enthusiasm \
#for learning my students learn in many different ways using all of our senses and multiple intell
igences i use a wide range\
#of techniques to help all my students succeed students in my class come from a variety of differe
nt backgrounds which makes\
#for wonderful sharing of experiences and cultures including native americans our school is a cari
ng community of successful \
#learners which can be seen through collaborative student project based learning in and out of the
classroom kindergarteners \
#in my class love to work with hands on materials and have many different opportunities to
practice a skill before it is\
#mastered having the social skills to work cooperatively with friends is a crucial aspect of the k
indergarten curriculum\
#montana is the perfect place to learn about agriculture and nutrition my students love to role pl
ay in our pretend kitchen\
#in the early childhood classroom i have had several kids ask me can we try cooking with real food
i will take their idea \
#and create common core cooking lessons where we learn important math and writing concepts while c
ooking delicious healthy \
#food for snack time my students will have a grounded appreciation for the work that went into mak
ing the food and knowledge \
#of where the ingredients came from as well as how it is healthy for their bodies this project wou
ld expand our learning of \
#nutrition and agricultural cooking recipes by having us peel our own apples to make homemade appl
esauce make our own bread \
#and mix up healthy plants from our classroom garden in the spring we will also create our own coo
kbooks to be printed and \
#shared with families students will gain math and literature skills as well as a life long enjoyme
nt for healthy cooking \
#nannan'
#ss = sid.polarity scores(for sentiment)
## The end=' ' is just to say that you want a space after the end of the statement instead of a ne
w line character.
#for k in ss:
   print('{0}: {1}, '.format(k, ss[k]), end='')
#for k in ss:
    print('{0}: {1}, '.format(k, ss[k]))
# we can use these 4 things as features/attributes (neg, neu, pos, compound)
# neg: 0.0, neu: 0.753, pos: 0.247, compound: 0.93
#print(type(ss))
#print(ss)
```

In [86]:

```
sentiment = sid.polarity_scores(sentance)
   preprocessed sentiments.append([sentiment['neg'], sentiment['pos'], sentiment['neu'],
[nltk data] Downloading package vader lexicon to
[nltk data]
              C:\Users\samar\AppData\Roaming\nltk data...
            Package vader lexicon is already up-to-date!
[nltk data]
100%|
                                                                              109248/109248
[06:13<00:00, 292.88it/s]
In [87]:
print(type(preprocessed sentiments))
print(preprocessed sentiments[1:5])
#print(preprocessed_sentiments([sentiment['neg']]))
print(sentiment['neg'])
<class 'list'>
[[0.037,\ 0.112,\ 0.851,\ 0.9267],\ [0.058,\ 0.179,\ 0.764,\ 0.995],\ [0.052,\ 0.214,\ 0.733,\ 0.9931],\ [0.013]
6, 0.087, 0.897, 0.9192]]
0.023
In [88]:
project_data[['neg', 'pos', 'neu', 'compound']] = pd.DataFrame(preprocessed_sentiments)
In [89]:
print(project data.columns.values)
project data['neg'].values
['Unnamed: 0' 'id' 'teacher_id' 'school_state'
 'project_submitted_datetime' 'project_title' 'project_essay_1'
 'project essay 2' 'project essay 3' 'project essay 4'
 'project_resource_summary' 'teacher_number_of_previously_posted_projects'
 'project is approved' 'clean categories' 'clean subcategories'
 'clean_grade' 'clean_tea_pfx' 'essay' 'neg' 'pos' 'neu' 'compound']
Out[89]:
array([0.008, 0.037, 0.058, ..., 0. , 0.013, 0.023])
Vectorizing Numerical features
In [90]:
price_data = resource_data.groupby('id').agg({'price':'sum', 'quantity':'sum'}).reset_index()
project_data = pd.merge(project_data, price_data, on='id', how='left')
print("Number of data points in train data", project_data.shape)
print('-'*50)
print("The attributes of data :", project data.columns.values)
Number of data points in train data (109248, 24)
```

Adding word count and length column as per suggestion 5

The attributes of data : ['Unnamed: 0' 'id' 'teacher_id' 'school_state'

 $\verb|'project_resource_summary'| | \verb|'teacher_number_of_previously_posted_projects'|$

'project_submitted_datetime' 'project_title' 'project_essay_1'

'project_is_approved' 'clean_categories' 'clean_subcategories' 'clean_grade' 'clean_tea_pfx' 'essay' 'neg' 'pos' 'neu' 'compound'

'project_essay_2' 'project_essay_3' 'project_essay_4'

'price' 'quantity']

```
In [91]:
```

```
project_data['essay_wc'] = preprocessed_essays_wc
project_data['essay_len'] = preprocessed_essays_len

project_data['title_wc'] = preprocessed_title_wc
project_data['title_len'] = preprocessed_title_len

project_data['prj_res_sum_wc'] = preprocessed_prj_sum_wc
project_data['prj_res_sum_len'] = preprocessed_prj_sum_len

print("Number of data points in train data", project_data.shape)
print('-'*50)
print("The attributes of data :", project_data.columns.values)
```

Assignment 5: Logistic Regression

- 1. [Task-1] Logistic Regression(either SGDClassifier with log loss, or LogisticRegression) on these feature sets
 - Set 1: categorical, numerical features + project_title(BOW) + preprocessed_eassay (`BOW with bi-grams` with
 `min df=10` and `max features=5000`)
 - Set 2: categorical, numerical features + project_title(TFIDF)+ preprocessed_eassay (`TFIDF with bi-grams` with `min_df=10` and `max_features=5000`)
 - Set 3: categorical, numerical features + project_title(AVG W2V)+ preprocessed_eassay (AVG W2V)
 - Set 4: categorical, numerical features + project_title(TFIDF W2V)+ preprocessed_essay (TFIDF W2V)
- 2. Hyper paramter tuning (find best hyper parameters corresponding the algorithm that you choose)
 - Find the best hyper parameter which will give the maximum AUC value
 - Find the best hyper paramter using k-fold cross validation or simple cross validation data
 - Use gridsearch cv or randomsearch cv or you can also write your own for loops to do this task of hyperparameter tuning
- 3. Representation of results
 - You need to plot the performance of model both on train data and cross validation data for each hyper parameter, like shown in the figure.
 - Once after you found the best hyper parameter, you need to train your model with it, and find the AUC on test data and plot the ROC curve on both train and test.
 - Along with plotting ROC curve, you need to print the <u>confusion matrix</u> with predicted and original labels of test data points. Please visualize your confusion matrices using <u>seaborn heatmaps</u>.
- 4. [Task-2] Apply Logistic Regression on the below feature set Set 5 by finding the best hyper parameter as suggested in step 2 and step 3.
- 5. Consider these set of features Set 5:
 - school_state : categorical data
 - clean_categories : categorical data
 - clean_subcategories : categorical data
 - project_grade_category :categorical data
 - teacher_prefix : categorical data
 - quantity: numerical data
 - teacher_number_of_previously_posted_projects : numerical data
 - price : numerical data
 - sentiment score's of each of the essay : numerical data
 - number of words in the title: numerical data

- Italian of troise in the title . Italianonous data

• <u>number of words in the combine essays</u>: numerical data

And apply the Logistic regression on these features by finding the best hyper paramter as suggested in step 2 and step 3

6. Conclusion

• You need to summarize the results at the end of the notebook, summarize it in the table format. To print out a table please refer to this prettytable library link

Note: Data Leakage

- 1. There will be an issue of data-leakage if you vectorize the entire data and then split it into train/cv/test.
- 2. To avoid the issue of data-leakage, make sure to split your data first and then vectorize it.
- 3. While vectorizing your data, apply the method fit_transform() on you train data, and apply the method transform() on cv/test data
- 4. For more details please go through this link.

2. Logistic Regression

```
In [92]:
```

```
##taking 50K datapoint
#project_data50K=project_data[:50000]
project_data100K=project_data[:100000]
X=project_data100K
#print(project_data50K.shape)
print(project_data100K.shape)
print(X.shape)
(100000, 30)
(100000, 30)
```

In [93]:

```
# makins Xi as 19 column matrix, where we create the modle and Yi as single colum matrix as a clas
s label.
#y = project data50K['project is approved'].values
#project data50K.drop(['project is approved'], axis=1, inplace=True)
#print(y.shape)
#project data50K.head(1)
y = project_data['project_is_approved'].values
project data.drop(['project is approved'], axis=1, inplace=True)
#print(y.shape)
project_data.head(1)
y100K=y[:100000]
y=y100K
#y = project_data['project_is_approved'].values
#project data.drop(['project is approved'], axis=1, inplace=True)
print(y.shape)
#project data.head(1)
```

(100000,)

In [94]:

```
#X = project_data50K
print(X.shape)
print(y.shape)
#X1K = project_data1K
#print(X1K.shape)
(100000, 30)
```

```
(100000, 30
```

2.1 Splitting data into Train and cross validation(or test): Stratified Sampling

```
In [47]:
```

```
# please write all the code with proper documentation, and proper titles for each subsection
# go through documentations and blogs before you start coding
# first figure out what to do, and then think about how to do.
# reading and understanding error messages will be very much helpfull in debugging your code
# when you plot any graph make sure you use
# a. Title, that describes your plot, this will be very helpful to the reader
# b. Legends if needed
# c. X-axis label
# d. Y-axis label
```

In [95]:

```
# train test split | https://scikit-
learn.org/stable/modules/generated/sklearn.model_selection.train_test_split.html
# spliting Xq and Yq in Train(further into Train and CV) and Test matrix
from sklearn.model_selection import train_test_split
X_train, X_test, Y_train, Y_test = train_test_split(X, y, test_size=0.33, stratify=y)
#X_train, X_cv, y_train, y_cv = train_test_split(X_train, y_train, test_size=0.33,
stratify=y_train)

print(X_train.shape, y_train.shape)
#print(X_cv.shape, y_cv.shape)
print(X_test.shape, y_test.shape)

print("="*100)

(67000, 30) (67000,)
(33000, 30) (33000,)
```

- N ►

2.1.1 Make Data Model Ready: encoding school_state categorical data

```
In [96]:
```

```
from sklearn.feature_extraction.text import CountVectorizer
vectorizer = CountVectorizer(min_df=10,ngram_range=(1,2), max_features=5000)
vectorizer.fit(X train['school state'].values) # fit has to happen only on train data
# we use the fitted CountVectorizer to convert the text to vector
X train state ohe = vectorizer.transform(X train['school state'].values)
#X cv state ohe = vectorizer.transform(X cv['school state'].values)
X test state ohe = vectorizer.transform(X test['school state'].values)
print("school state After vectorizations")
print(X train state_ohe.shape, y_train.shape)
#print(X cv state ohe.shape, y cv.shape)
print(X test state ohe.shape, y test.shape)
print(vectorizer.get feature names())
print("="*100)
school state After vectorizations
(67000, 51) (67000,)
(33000, 51) (33000,)
['ak', 'al', 'ar', 'az', 'ca', 'co', 'ct', 'dc', 'de', 'fl', 'ga', 'hi', 'ia', 'id', 'il', 'in', 'k
s', 'ky', 'la', 'ma', 'md', 'me', 'mi', 'mn', 'mo', 'ms', 'mt', 'nc', 'nd', 'ne', 'nh', 'nj', 'nm',
'nv', 'ny', 'oh', 'ok', 'or', 'pa', 'ri', 'sc', 'sd', 'tn', 'tx', 'ut', 'va', 'vt', 'wa', 'wi', 'wv
', 'wy']
_____
```

2.1.2 Make Data Model Ready: encoding clean_categories

```
from sklearn.feature_extraction.text import CountVectorizer
#vectorizer = CountVectorizer(min df=10,ngram range=(1,2), max features=5000)
vectorizer = CountVectorizer(vocabulary =list(sorted cat dict.keys()), lowercase =False, binary=True
vectorizer.fit(X train['clean categories'].values) # fit has to happen only on train data
# we use the fitted CountVectorizer to convert the text to vector
X train clean ohe = vectorizer.transform(X train['clean categories'].values)
\#X cv clean ohe = vectorizer.transform(X \overline{\text{cv}}['\text{clean categories'}].values)
X_test_clean_ohe = vectorizer.transform(X_test['clean_categories'].values)
print("clean categories After vectorizations")
print(X_train_clean_ohe.shape, y_train.shape)
#print(X_cv_clean_ohe.shape, y_cv.shape)
print(X_test_clean_ohe.shape, y_test.shape)
print(vectorizer.get feature names())
print("="*100)
clean categories After vectorizations
(67000, 9) (67000,)
(33000, 9) (33000,)
['Warmth', 'Care_Hunger', 'History_Civics', 'Music_Arts', 'AppliedLearning', 'SpecialNeeds',
'Health Sports', 'Math Science', 'Literacy Language']
```

2.1.3 Make Data Model Ready: encoding clean_subcategories

```
In [98]:
```

```
from sklearn.feature extraction.text import CountVectorizer
vectorizer = CountVectorizer(vocabulary =list(sorted sub cat dict.keys()), lowercase =False, binary=
True)
vectorizer.fit(X_train['clean_subcategories'].values) # fit has to happen only on train data
# we use the fitted CountVectorizer to convert the text to vector
X train cleanSub ohe = vectorizer.transform(X train['clean subcategories'].values)
#X cv cleanSub ohe = vectorizer.transform(X cv['clean subcategories'].values)
X_test_cleanSub_ohe = vectorizer.transform(X_test['clean_subcategories'].values)
print("clean subcategories After vectorizations")
print(X_train_cleanSub_ohe.shape, y_train.shape)
#print(X cv cleanSub ohe.shape, y cv.shape)
print(X_test_cleanSub_ohe.shape, y_test.shape)
#print(vectorizer.get feature names())
print("="*100)
clean subcategories After vectorizations
(67000, 30) (67000,)
(33000, 30) (33000,)
```

2.1.4 Make Data Model Ready: encoding project_grade_category

In [99]:

```
from sklearn.feature_extraction.text import CountVectorizer
vectorizer = CountVectorizer(vocabulary = list(sorted_prj_grade_cat_dict.keys()), lowercase = False, b
inary=True)
vectorizer.fit(X_train['clean_grade'].values) # fit has to happen only on train data

# we use the fitted CountVectorizer to convert the text to vector
X_train_grade_ohe = vectorizer.transform(X_train['clean_grade'].values)

#X_cv_grade_ohe = vectorizer.transform(X_cv['clean_grade'].values)
X_test_grade_ohe = vectorizer.transform(X_test['clean_grade'].values)

print("project_grade_category After vectorizations")
print(X_train_grade_ohe.shape, y_train.shape)
#print(X_cv_grade_ohe.shape, y_cv.shape)
print(X_test_grade_ohe.shape, y_test.shape)
```

```
print(vectorizer.get feature_names())
print("="*100)
project_grade_category After vectorizations
(67000, 4) (67000,)
(33000, 4) (33000,)
['9-12', '6-8', '3-5', 'PreK-2']
```

2.1.5 Make Data Model Ready: encoding teacher prefix

```
In [100]:
from sklearn.feature extraction.text import CountVectorizer
vectorizer = CountVectorizer(vocabulary =list(sorted tea pfx cat dict.keys()), lowercase =False, bin
\# https://stackoverflow.com/questions/52736900/how-to-solve-the-attribute-error-float-object-has-no-like theorem and the solve-the-attribute-error-float-object-has-no-like the-attribute-error-float-object-has-no-like the-attribute-error-float-has-no-like the-attribute-error-float-has-no-like the-attribute-has-no-like the-attribute-has-no-like the-attribute-has-no-like the-attribute-has-no-like the-attrib
 -attribute-split-in-pyth
vectorizer.fit(X_train['clean_tea_pfx'].astype(str).values) # fit has to happen only on train data
 # we use the fitted CountVectorizer to convert the text to vector
X train teacher ohe = vectorizer.transform(X train['clean tea pfx'].astype(str).values)
 #X cv teacher one = vectorizer.transform(X cv['clean tea pfx'].astype(str).values)
X test teacher ohe = vectorizer.transform(X test['clean tea pfx'].astype(str).values)
print("teacher_prefix After vectorizations")
print(X train teacher ohe.shape, y train.shape)
#print(X_cv_teacher_ohe.shape, y_cv.shape)
print(X test teacher ohe.shape, y test.shape)
print(vectorizer.get feature names())
print("="*100)
teacher prefix After vectorizations
(67000, 5) (67000,)
(33000, 5) (33000,)
['Dr', 'Teacher', 'Mr', 'Ms', 'Mrs']
```

2.1.6 Make Data Model Ready: encoding project resource summary

```
In [101]:
```

```
vectorizer = CountVectorizer(min df=10,ngram range=(1,2))
vectorizer.fit(X_train['project_resource_summary'].values) # fit has to happen only on train data
# we use the fitted CountVectorizer to convert the text to vector
X_train_prjResSum_ohe = vectorizer.transform(X_train['project_resource_summary'].values)
#X_cv_prjResSum_ohe = vectorizer.transform(X_cv['project_resource_summary'].values)
X test prjResSum ohe = vectorizer.transform(X test['project resource summary'].values)
print("project resource summary After vectorizations")
print(X train prjResSum ohe.shape, y train.shape)
#print(X cv prjResSum ohe.shape, y cv.shape)
print(X test_prjResSum_ohe.shape, y_test.shape)
#print(vectorizer.get_feature_names())
print("="*100)
project resource summary After vectorizations
(67000, 18601) (\overline{6}7000,)
(33000, 18601) (33000,)
```

2.2 Make Data Model Ready: encoding numerical, categorical features

```
# please write all the code with proper documentation, and proper titles for each subsection
# go through documentations and blogs before you start coding
# first figure out what to do, and then think about how to do.
# reading and understanding error messages will be very much helpfull in debugging your code
# make sure you featurize train and test data separatly

# when you plot any graph make sure you use
# a. Title, that describes your plot, this will be very helpful to the reader
# b. Legends if needed
# c. X-axis label
# d. Y-axis label
```

2.2.1 Make Data Model Ready: encoding numerical | quantity

In [102]:

```
from sklearn.preprocessing import Normalizer
normalizer = Normalizer()
# normalizer.fit(X train['price'].values)
# this will rise an error Expected 2D array, got 1D array instead:
# array=[105.22 215.96 96.01 ... 368.98 80.53 709.67].
# Reshape your data either using
# array.reshape(-1, 1) if your data has a single feature
\# array.reshape(1, -1) if it contains a single sample.
normalizer.fit(X_train['quantity'].values.reshape(-1,1))
X train quantity norm = normalizer.transform(X train['quantity'].values.reshape(-1,1))
#X cv quantity norm = normalizer.transform(X cv['quantity'].values.reshape(-1,1))
X test quantity norm = normalizer.transform(X test['quantity'].values.reshape(-1,1))
print("quantity After vectorizations")
print(X_train_quantity_norm.shape, y_train.shape)
#print(X cv quantity norm.shape, y cv.shape)
print(X test quantity norm.shape, y test.shape)
print("="*100)
quantity After vectorizations
(67000, 1) (67000,)
(33000, 1) (33000,)
```

2.2.2 Make Data Model Ready: encoding numerical| teacher_number_of_previously_posted_projects

```
In [103]:
```

```
from sklearn.preprocessing import Normalizer
normalizer = Normalizer()
# normalizer.fit(X train['price'].values)
# this will rise an error Expected 2D array, got 1D array instead:
# array=[105.22 215.96 96.01 ... 368.98 80.53 709.67].
# Reshape your data either using
# array.reshape(-1, 1) if your data has a single feature
# array.reshape(1, -1) if it contains a single sample.
normalizer.fit(X train['teacher number of previously posted projects'].values.reshape(-1,1))
X train TprevPrj norm =
normalizer.transform(X train['teacher number of previously posted projects'].values.reshape(-1,1))
#X cv TprevPrj norm =
normalizer.transform(X cv['teacher number of previously posted projects'].values.reshape(-1,1))
X test TprevPrj norm = normalizer.transform(X test['teacher number of previously posted projects']
.values.reshape(-1,1))
print("teacher_number_of_previously_posted_projects After vectorizations")
print(X_train_TprevPrj_norm.shape, y_train.shape)
#print(X_cv_TprevPrj_norm.shape, y_cv.shape)
print(X_test_TprevPrj_norm.shape, y_test.shape)
print("="*100)
```

2.2.3 Make Data Model Ready: encoding numerical | price

```
In [104]:
```

```
from sklearn.preprocessing import Normalizer
normalizer = Normalizer()
# normalizer.fit(X_train['price'].values)
# this will rise an error Expected 2D array, got 1D array instead:
# array=[105.22 215.96 96.01 ... 368.98 80.53 709.67].
# Reshape your data either using
# array.reshape(-1, 1) if your data has a single feature
# array.reshape(1, -1) if it contains a single sample.
normalizer.fit(X train['price'].values.reshape(-1,1))
X_train_price_norm = normalizer.transform(X_train['price'].values.reshape(-1,1))
#X_cv_price_norm = normalizer.transform(X_cv['price'].values.reshape(-1,1))
X_test_price_norm = normalizer.transform(X_test['price'].values.reshape(-1,1))
print("Price After vectorizations")
print(X_train_price_norm.shape, y_train.shape)
#print(X_cv_price_norm.shape, y_cv.shape)
print(X_test_price_norm.shape, y_test.shape)
print("="*100)
Price After vectorizations
(67000, 1) (67000,)
(33000, 1) (33000,)
```

2.2.4 Make Data Model Ready: encoding numerical | sentimental score

2.2.4.1 Make Data Model Ready: encoding numerical | sentimental score | neg

In [105]:

```
from sklearn.preprocessing import Normalizer
normalizer = Normalizer()
# normalizer.fit(X train['price'].values)
# this will rise an error Expected 2D array, got 1D array instead:
# array=[105.22 215.96 96.01 ... 368.98 80.53 709.67].
# Reshape your data either using
# array.reshape(-1, 1) if your data has a single feature
# array.reshape(1, -1) if it contains a single sample.
normalizer.fit(X train['neg'].values.reshape(-1,1))
X train neg norm = normalizer.transform(X train['neg'].values.reshape(-1,1))
#X_cv_neg_norm = normalizer.transform(X_cv['neg'].values.reshape(-1,1))
X_test_neg_norm = normalizer.transform(X_test['neg'].values.reshape(-1,1))
print("neg After vectorizations")
print(X_train_neg_norm.shape, y_train.shape)
#print(X cv neg_norm.shape, y_cv.shape)
print(X test neg norm.shape, y test.shape)
print("="*100)
neg After vectorizations
(67000, 1) (67000,)
(33000, 1) (33000,)
```

2.2.4.2 Make Data Model Ready: encoding numerical | sentimental score | pos

In [106]:

```
from sklearn.preprocessing import Normalizer
normalizer = Normalizer()
# normalizer.fit(X train['price'].values)
# this will rise an error Expected 2D array, got 1D array instead:
# array=[105.22 215.96 96.01 ... 368.98 80.53 709.67].
# Reshape your data either using
# array.reshape(-1, 1) if your data has a single feature
# array.reshape(1, -1) if it contains a single sample.
normalizer.fit(X train['pos'].values.reshape(-1,1))
X_train_pos_norm = normalizer.transform(X_train['pos'].values.reshape(-1,1))
#X cv pos norm = normalizer.transform(X cv['pos'].values.reshape(-1,1))
X_test_pos_norm = normalizer.transform(X_test['pos'].values.reshape(-1,1))
print("pos After vectorizations")
print(X_train_pos_norm.shape, y_train.shape)
#print(X cv_pos_norm.shape, y_cv.shape)
print(X test pos norm.shape, y test.shape)
print("="*100)
pos After vectorizations
(67000, 1) (67000,)
(33000, 1) (33000,)
```

2.2.4.3 Make Data Model Ready: encoding numerical | sentimental score | neu

In [107]:

```
from sklearn.preprocessing import Normalizer
normalizer = Normalizer()
# normalizer.fit(X train['price'].values)
# this will rise an error Expected 2D array, got 1D array instead:
# array=[105.22 215.96 96.01 ... 368.98 80.53 709.67].
# Reshape your data either using
# array.reshape(-1, 1) if your data has a single feature
# array.reshape(1, -1) if it contains a single sample.
normalizer.fit(X_train['neu'].values.reshape(-1,1))
X_train_neu_norm = normalizer.transform(X_train['neu'].values.reshape(-1,1))
#X cv neu norm = normalizer.transform(X cv['neu'].values.reshape(-1,1))
X test neu norm = normalizer.transform(X test['neu'].values.reshape(-1,1))
print("neu After vectorizations")
print(X train neu norm.shape, y train.shape)
#print(X_cv_neu_norm.shape, y_cv.shape)
print(X test neu_norm.shape, y_test.shape)
print("="*100)
neu After vectorizations
(67000, 1) (67000,)
(33000, 1) (33000,)
```

2.2.4.4 Make Data Model Ready: encoding numerical | sentimental score | compound

In [108]:

```
from sklearn.preprocessing import Normalizer
normalizer = Normalizer()
# normalizer.fit(X_train['price'].values)
# this will rise an error Expected 2D array, got 1D array instead:
# array=[105.22 215.96 96.01 ... 368.98 80.53 709.67].
# Reshape your data either using
# array.reshape(-1, 1) if your data has a single feature
```

2.2.5 Make Data Model Ready: encoding numerical | number of words in the title

In [109]:

```
from sklearn.preprocessing import Normalizer
normalizer = Normalizer()
# normalizer.fit(X train['price'].values)
# this will rise an error Expected 2D array, got 1D array instead:
# array=[105.22 215.96 96.01 ... 368.98 80.53 709.67].
# Reshape your data either using
# array.reshape(-1, 1) if your data has a single feature
# array.reshape(1, -1) if it contains a single sample.
normalizer.fit(X_train['title_wc'].values.reshape(-1,1))
X train title wc norm = normalizer.transform(X train['title wc'].values.reshape(-1,1))
\#X\_cv\_title\_wc\_norm = normalizer.transform(X\_cv['title\_wc'].values.reshape(-1,1))
X test title wc norm = normalizer.transform(X test['title wc'].values.reshape(-1,1))
print("title wc After vectorizations")
print (X train title wc norm.shape, y train.shape)
#print(X_cv_title_wc_norm.shape, y_cv.shape)
print(X test title wc norm.shape, y test.shape)
print("="*100)
title wc After vectorizations
(67000, 1) (67000,)
(33000, 1) (33000,)
```

2.2.6 Make Data Model Ready: encoding numerical | number of words in the essay

```
In [110]:
```

```
from sklearn.preprocessing import Normalizer
normalizer = Normalizer()
# normalizer.fit(X_train['price'].values)
# this will rise an error Expected 2D array, got 1D array instead:
# array=[105.22 215.96 96.01 ... 368.98 80.53 709.67].
# Reshape your data either using
# array.reshape(-1, 1) if your data has a single feature
# array.reshape(1, -1) if it contains a single sample.
normalizer.fit(X_train['essay_wc'].values.reshape(-1,1))

X_train_essay_wc_norm = normalizer.transform(X_train['essay_wc'].values.reshape(-1,1))
#X_cv_essay_wc_norm = normalizer.transform(X_cv['essay_wc'].values.reshape(-1,1))
X_test_essay_wc_norm = normalizer.transform(X_test['essay_wc'].values.reshape(-1,1))
print("essay_wc After vectorizations")
```

2.3 Make Data Model Ready: encoding eassay, and project title

```
In [111]:
```

```
# please write all the code with proper documentation, and proper titles for each subsection
# go through documentations and blogs before you start coding
# first figure out what to do, and then think about how to do.
# reading and understanding error messages will be very much helpfull in debugging your code
# make sure you featurize train and test data separatly

# when you plot any graph make sure you use
# a. Title, that describes your plot, this will be very helpful to the reader
# b. Legends if needed
# c. X-axis label
# d. Y-axis label
```

2.3.1 Make Data Model Ready: project_essay | BOW

In [112]:

```
from sklearn.feature_extraction.text import CountVectorizer
# categorical, numerical features + project_title(BOW) + preprocessed_eassay
\# (BOW with bi-grams with min df=10 and max features=5000)
vectorizer = CountVectorizer(min df=10,ngram range=(1,2), max features=5000)
vectorizer.fit(X_train['essay'].values) # fit has to happen only on train data
# we use the fitted CountVectorizer to convert the text to vector
X_train_essay_bow = vectorizer.transform(X_train['essay'].values)
#X cv essay bow = vectorizer.transform(X cv['essay'].values)
X test essay bow = vectorizer.transform(X test['essay'].values)
print("Essay After vectorizations")
print(X_train_essay_bow.shape, y_train.shape)
#print(X cv essay_bow.shape, y_cv.shape)
print(X_test_essay_bow.shape, y_test.shape)
print("="*100)
Essay After vectorizations
(67000, 5000) (67000,)
(33000, 5000) (33000,)
```

2.3.2 Make Data Model Ready: project_title | BOW

```
In [113]:
```

```
vectorizer = CountVectorizer()
# categorical, numerical features + project_title(BOW) + preprocessed_eassay
# (BOW with bi-grams with min_df=10 and max_features=5000)
vectorizer = CountVectorizer(min_df=10,ngram_range=(1,2), max_features=5000)
vectorizer.fit(X_train['project_title'].values) # fit has to happen only on train data

# we use the fitted CountVectorizer to convert the text to vector
X_train_title_bow = vectorizer.transform(X_train['project_title'].values)
#X_cv_title_bow = vectorizer.transform(X_cv['project_title'].values)
X_test_title_bow = vectorizer.transform(X_test['project_title'].values)
```

```
print("project_title After vectorizations")
print(X_train_title_bow.shape, y_train.shape)
#print(X cv_title_bow.shape, y_cv.shape)
print(X_test_title_bow.shape, y_test.shape)
#print(vectorizer.get_feature_names())
print("="*100)
project title After vectorizations
(67000, 5000) (67000,)
(33000, 5000) (33000,)
```

2.3.3 Make Data Model Ready: project_essay | TFIDF

In [114]:

```
from sklearn.feature extraction.text import TfidfVectorizer
# categorical, numerical features + project_title(BOW) + preprocessed_eassay
# (TFIDF with bi-grams with min_df=10 and max_features=5000)
Tfidf vectorizer = TfidfVectorizer(min df=10,ngram range=(1,2), max features=5000)
Tfidf vectorizer.fit(X train['essay'].values)
X train text tfidf = Tfidf vectorizer.transform(X train['essay'].values)
#X cv text tfidf = Tfidf vectorizer.transform(X cv['essay'].values)
X_test_text_tfidf = Tfidf_vectorizer.transform(X_test['essay'].values)
##print("Shape of matrix after one hot encodig ",text tfidf.shape)
print("Essay After vectorizations")
print(X train text tfidf.shape, y train.shape)
#print(X_cv_text_tfidf.shape, y_cv.shape)
print(X_test_text_tfidf.shape, y_test.shape)
#print(Tfidf vectorizer.get feature names())
print("="*100)
Essay After vectorizations
(67000, 5000) (67000,)
```

```
(33000, 5000) (33000,)
```

.....▶

2.3.4 Make Data Model Ready: project_title | TFIDF

In [115]:

```
from sklearn.feature extraction.text import TfidfVectorizer
# categorical, numerical features + project title(BOW) + preprocessed eassay
# (TFIDF with bi-grams with min df=10 and max features=5000)
Tfidf vectorizer = TfidfVectorizer(min df=10,ngram range=(1,2), max features=5000)
Tfidf_vectorizer.fit(X_train['project_title'].values)
X train title tfidf = Tfidf vectorizer.transform(X train['project title'].values)
#X cv title tfidf = Tfidf vectorizer.transform(X cv['project title'].values)
X_test_title_tfidf = Tfidf_vectorizer.transform(X_test['project_title'].values)
##print("Shape of matrix after one hot encodig ",text tfidf.shape)
print("project title After vectorizations")
print(X train title tfidf.shape, y train.shape)
#print(X_cv_title_tfidf.shape, y_cv.shape)
print(X test title tfidf.shape, y test.shape)
#print(Tfidf vectorizer.get_feature_names())
print("="*100)
project_title After vectorizations
(67000, 5000) (67000,)
(33000, 5000) (33000,)
```

2.3.5 Make Data Model Ready: project_essay | AVG W2V

```
In [116]:
```

```
# average Word2Vec for Train Essay
# compute average word2vec for each review.
X_{train}_{say} = []; \# the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(X_train['essay'].values): # for each review/sentence
   vector = np.zeros(300) # as word vectors are of zero length
    cnt words =0; # num of words with a valid vector in the sentence/review
    for word in sentence.split(): # for each word in a review/sentence
       if word in glove words:
           vector += model[word]
            cnt words += 1
    if cnt words != 0:
       vector /= cnt_words
    X train essay avg w2v.append(vector)
print(len(X_train_essay_avg_w2v))
print(len(X train_essay_avg_w2v[0]))
# stronging variables into pickle files python: http://www.jessicayung.com/how-to-use-pickle-to-sa
ve-and-load-variables-in-python/
import pickle
with open('X_train_essay_avg_w2v', 'wb') as f:
   pickle.dump(X train essay avg w2v, f)
                                                                            67000/67000
[00:23<00:00, 2876.98it/s]
67000
300
```

In [117]:

In [118]:

```
# average Word2Vec for Test Essay
# compute average word2vec for each review.
X_{\text{test\_essay\_avg\_w2v}} = []; \# the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(X_test['essay'].values): # for each review/sentence
   vector = np.zeros(300) # as word vectors are of zero length
   cnt words =0; # num of words with a valid vector in the sentence/review
    for word in sentence.split(): # for each word in a review/sentence
        if word in glove words:
            vector += model[word]
            cnt words += 1
    if cnt words != 0:
        vector /= cnt words
    X test essay avg w2v.append(vector)
print(len(X test essay avg w2v))
print(len(X_test_essay_avg_w2v[0]))
                                                                                | 33000/33000
[00:12<00:00, 2593.91it/s]
```

ろろししし

2.3.6 Make Data Model Ready: project_title | AVG W2V

```
In [119]:
# average Word2Vec for Train Title
# compute average word2vec for each review.
X train title avg w2v = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(X_train['project_title'].values): # for each review/sentence
    vector = np.zeros(300) # as word vectors are of zero length
    cnt words =0; # num of words with a valid vector in the sentence/review
    for word in sentence.split(): # for each word in a review/sentence
        if word in glove words:
            vector += model[word]
            cnt words += 1
    if cnt words != 0:
        vector /= cnt words
    X_train_title_avg_w2v.append(vector)
print(len(X_train_title_avg_w2v))
print(len(X_train_title_avg_w2v[0]))
100%|
                                                                              | 67000/67000
[00:00<00:00, 133454.07it/s]
67000
300
In [120]:
# average Word2Vec for Test Essay
# compute average word2vec for each review.
X_{\text{test\_title\_avg\_w2v}} = []; \# the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(X test['project title'].values): # for each review/sentence
   vector = np.zeros(300) # as word vectors are of zero length
    cnt words =0; # num of words with a valid vector in the sentence/review
    for word in sentence.split(): # for each word in a review/sentence
        if word in glove words:
            vector += model[word]
            cnt words += 1
    if cnt_words != 0:
        vector /= cnt words
    X test title avg w2v.append(vector)
print(len(X test title avg w2v))
print(len(X_test_title_avg_w2v[0]))
                                                                               | 33000/33000
[00:00<00:00, 152776.37it/s]
33000
```

2.3.7 Make Data Model Ready: project essay | TFIDF W2V

```
In [121]:
```

300

```
# S = ["abc def pqr", "def def def abc", "pqr pqr def"]
Tr_tfidf_model_essay = TfidfVectorizer()
Tr_tfidf_model_essay.fit(X_train['essay'].values)
# we are converting a dictionary with word as a key, and the idf as a value
dictionary = dict(zip(Tr_tfidf_model_essay.get_feature_names(), list(Tr_tfidf_model_essay.idf_)))
tr_essay_tfidf_words = set(Tr_tfidf_model_essay.get_feature_names())
```

```
In [122]:
```

```
# compute average word2vec for each review.
tr_tfidf_w2v_essay_vectors = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(X_train['essay'].values): # for each review/sentence
    vector = np.zeros(300) # as word vectors are of zero length
    tf idf weight =0; # num of words with a valid vector in the sentence/review
    for word in sentence.split(): # for each word in a review/sentence
        if (word in glove words) and (word in tr essay tfidf words):
            vec = model[word] # getting the vector for each word
            # here we are multiplying idf value(dictionary[word]) and the tf
value((sentence.count(word)/len(sentence.split())))
            tf idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # getting the tf
idf value for each word
            vector += (vec * tf_idf) # calculating tfidf weighted w2v
            tf idf weight += tf idf
    if tf_idf_weight != 0:
        vector /= tf idf weight
    tr_tfidf_w2v_essay_vectors.append(vector)
print(len(tr_tfidf_w2v_essay_vectors))
print(len(tr tfidf w2v essay vectors[0]))
100%|
                                                                        67000/67000 [04:
21<00:00, 256.06it/s]
67000
300
In [123]:
\# S = ["abc def pqr", "def def def abc", "pqr pqr def"]
te_tfidf_model_essay = TfidfVectorizer()
te tfidf model essay.fit(X test['essay'].values)
# we are converting a dictionary with word as a key, and the idf as a value
dictionary = dict(zip(te_tfidf_model_essay.get_feature_names(), list(te_tfidf_model_essay.idf_)))
te tfidf model essay = set(te tfidf model essay.get feature names())
In [124]:
# TFIDF weighted Word2Vec for test essay
# compute average word2vec for each review.
\texttt{te\_tfidf\_w2v\_essay\_vectors} = \texttt{[]; \# the avg-w2v for each sentence/review is stored in this list}
for sentence in tqdm (X test['essay'].values): # for each review/sentence
    vector = np.zeros(300) # as word vectors are of zero length
    \verb|tf_idf_weight = 0|; \# num \ of \ words \ with \ a \ valid \ vector \ in \ the \ sentence/review
    for word in sentence.split(): # for each word in a review/sentence
        if (word in glove words) and (word in te tfidf model essay):
            vec = model[word] # getting the vector for each word
            # here we are multiplying idf value(dictionary[word]) and the tf
value((sentence.count(word)/len(sentence.split())))
            tf idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # getting the tf
idf value for each word
            vector += (vec * tf idf) # calculating tfidf weighted w2v
            tf idf weight += tf idf
    if tf idf weight != 0:
       vector /= tf_idf_weight
```

33000/33000 [02:

33000 300

100%|

39<00:00, 207.42it/s]

2.3.8 Make Data Model Ready: project title | TFIDF W2V

te tfidf w2v essay vectors.append(vector)

print(len(te_tfidf_w2v_essay_vectors))
print(len(te_tfidf_w2v_essay_vectors[0]))

In [126]:

```
Tr tfidf model title = TfidfVectorizer()
Tr tfidf model title.fit(X train['project title'].values)
# we are converting a dictionary with word as a key, and the idf as a value
dictionary = dict(zip(Tr tfidf model title.get feature names(), list(Tr tfidf model title.idf )))
Tr tfidf model title = set(Tr tfidf model title.get feature names())
In [127]:
# TFIDF weighted Word2Vec for train title
# compute average word2vec for each review.
tr tfidf w2v title vectors = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(X_train['project_title'].values): # for each review/sentence
    vector = np.zeros(300) # as word vectors are of zero length
    tf idf weight =0; # num of words with a valid vector in the sentence/review
    for word in sentence.split(): # for each word in a review/sentence
        if (word in glove words) and (word in Tr tfidf model title):
            vec = model[word] # getting the vector for each word
            # here we are multiplying idf value(dictionary[word]) and the tf
value((sentence.count(word)/len(sentence.split())))
            tf_idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # getting the tf
idf value for each word
            vector += (vec * tf idf) # calculating tfidf weighted w2v
            tf_idf_weight += tf_idf
    if tf idf weight != 0:
        vector /= tf idf weight
    tr tfidf w2v title vectors.append(vector)
print(len(tr tfidf w2v title vectors))
print(len(tr_tfidf_w2v_title_vectors[0]))
                                                                          | 67000/67000
100%|
[00:00<00:00, 87991.64it/s]
67000
300
In [128]:
\# S = ["abc def pqr", "def def def abc", "pqr pqr def"]
te tfidf model title = TfidfVectorizer()
te_tfidf_model_title.fit(X_test['project_title'].values)
# we are converting a dictionary with word as a key, and the idf as a value
dictionary = dict(zip(te tfidf model title.get feature names(), list(te tfidf model title.idf )))
te_tfidf_model_title = set(te_tfidf_model_title.get_feature_names())
In [129]:
# TFIDF weighted Word2Vec for test title
# compute average word2vec for each review.
te_tfidf_w2v_title_vectors = []; # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(X_test['project_title'].values): # for each review/sentence
    vector = np.zeros(300) # as word vectors are of zero length
    tf idf weight =0; # num of words with a valid vector in the sentence/review
    for word in sentence.split(): # for each word in a review/sentence
        if (word in glove words) and (word in te tfidf model title):
            vec = model[word] # getting the vector for each word
            # here we are multiplying idf value(dictionary[word]) and the tf
value((sentence.count(word)/len(sentence.split())))
           tf_idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # getting the tf
idf value for each word
            vector += (vec * tf idf) # calculating tfidf weighted w2v
            tf idf weight += tf_idf
    if tf idf weight != 0:
        vector /= tf_idf_weight
    te_tfidf_w2v_title_vectors.append(vector)
print(len(te_tfidf_w2v_title_vectors))
print(len(te tfidf w2v title vectors[0]))
                                                                      | 33000/33000
100%1
```

S = ["abc def pgr", "def def def abc", "pgr pgr def"]

[00:00<00:00, 79731.26it/s]

2.4 Appling Logistic Regression on different kind of featurization as mentioned in the instructions

Apply Logistic Regression on different kind of featurization as mentioned in the instructions For Every model that you work on make sure you do the step 2 and step 3 of instrucations

In [130]:

```
# please write all the code with proper documentation, and proper titles for each subsection
# go through documentations and blogs before you start coding
# first figure out what to do, and then think about how to do.
# reading and understanding error messages will be very much helpfull in debugging your code
# when you plot any graph make sure you use
# a. Title, that describes your plot, this will be very helpful to the reader
# b. Legends if needed
# c. X-axis label
# d. Y-axis label
```

1. [Task-1] Logistic Regression(either SGDClassifier with log loss, or LogisticRegression) on these feature sets

• Set 1: categorical, numerical features + project_title(BOW) + preprocessed_eassay (`BOW with bi-grams` with `min df=10` and `max features=5000`)

In [131]:

```
# merge two sparse matrices: https://stackoverflow.com/a/19710648/4084039
from scipy.sparse import hstack
X tr bow = hstack((X train_essay_bow, X_train_title_bow, X_train_state_ohe, X_train_clean_ohe,
X_train_cleanSub_ohe, X_train_grade_ohe, X_train_teacher_ohe, X_train_prjResSum_ohe,
X_train_quantity_norm, X_train_TprevPrj_norm, X_train_price_norm)).tocsr()
X_te_bow = hstack((X_test_essay_bow, X_test_title_bow, X_test_state_ohe, X_test_clean_ohe, X_test_cleanSub_ohe, X_test_grade_ohe, X_test_teacher_ohe, X_test_prjResSum_ohe, X_test_quantity_norm,
X_test_TprevPrj_norm, X_test_price_norm)).tocsr()
print("Final Data matrix | BOW")
print(X_tr_bow.shape, y_train.shape)
print(X_te_bow.shape, y_test.shape)
print("="*100)

Final Data matrix | BOW
(67000, 28703) (67000,)
(33000, 28703) (33000,)

***This important is a comparation of the print is a comparation of the
```

In [132]:

```
#c_range=[10**-5, 10**-4, 10**-2, 10**0, 10**2, 10**4, 10**5]
#param_grid=dict(C=c_range)
#print(type(c_range))
#print(c_range)
#print(type(param_grid))
#print(param_grid)
```

In [133]:

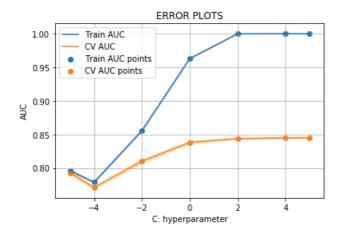
```
#code source:
http://occam.olin.edu/sites/default/files/DataScienceMaterials/machine_learning_lecture_2/Machine%2
rning%20Lecture%202.html
from sklearn.model_selection import train_test_split
#from sklearn.grid_search import GridSearchCV
from sklearn.model_selection import GridSearchCV
from sklearn.datasets import *
from sklearn.linear_model import LogisticRegression
```

In [135]:

0.8450351607083686

```
train bow auc= model.cv results ['mean train score']
train_bow_auc_std= model.cv_results_['std_train_score']
cv_bow_auc = model.cv_results_['mean_test_score']
cv bow auc std= model.cv_results_['std_test_score']
CC = []
from math import log
\#alpha = [log(x) for x in a]
CC = [np.log10(x) \text{ for } x \text{ in } c \text{ range}]
print(CC)
plt.plot(CC, train bow auc, label='Train AUC')
# this code is copied from here: https://stackoverflow.com/a/48803361/4084039
plt.gca().fill between(CC, train bow auc - train bow auc std, train bow auc +
train bow auc std,alpha=0.2,color='darkblue')
plt.plot(CC, cv bow auc, label='CV AUC')
# this code is copied from here: https://stackoverflow.com/a/48803361/4084039
plt.gca().fill between(CC,cv bow auc - cv bow auc std,cv bow auc + cv bow auc std,alpha=0.2,color='
darkorange')
plt.scatter(CC, train_bow_auc, label='Train AUC points')
plt.scatter(CC, cv bow auc, label='CV AUC points')
plt.legend()
plt.xlabel("C: hyperparameter")
plt.ylabel("AUC")
plt.title("ERROR PLOTS")
plt.grid(True)
plt.show()
```

[-5.0, -4.0, -2.0, 0.0, 2.0, 4.0, 5.0]

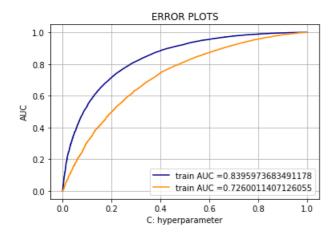


In [136]:

```
best_tuned_parameters = [{'C': [U.U1]}]
```

```
In [138]:
```

```
# https://scikit-
learn.org/stable/modules/generated/sklearn.metrics.roc curve.html#sklearn.metrics.roc curve
from sklearn.metrics import roc_curve, auc
model = GridSearchCV(LogisticRegression(), best tuned parameters)
model.fit(X tr bow, y train)
# roc auc score(y true, y score) the 2nd parameter should be probability estimates of the positive
class
# not the predicted outputs
y train bow pred = model.predict proba(X tr bow)[:,1]
y test bow pred = model.predict proba(X te bow)[:,1]
print(model.best estimator )
print(model.score(X_te_bow, y_test))
train fpr, train tpr, tr thresholds = roc curve (y train, y train bow pred)
test_fpr, test_tpr, te_thresholds = roc_curve(y_test, y_test_bow_pred)
plt.plot(train fpr, train tpr, label="train AUC ="+str(auc(train fpr, train tpr)),color='darkblue'
plt.plot(test fpr, test tpr, label="train AUC ="+str(auc(test fpr, test tpr)),color='darkorange')
plt.legend()
plt.xlabel("C: hyperparameter")
plt.ylabel("AUC")
plt.title("ERROR PLOTS")
plt.grid(True)
plt.show()
```



When class_weight=Auto

In [176]:

```
# Please compare all your models using Prettytable library
from prettytable import PrettyTable

x = PrettyTable()

x.field_names = ["Vectorizer", "Algorithm", "Hyper parameter", "Train AUC", "Test AUC"]

x.add_row(["BOW", "LogisticRegression", 0.00001, 0.57485, 0.56908 ])

x.add_row(["BOW", "LogisticRegression", 0.0001, 0.66466, 0.639418 ])

x.add_row(["BOW", "LogisticRegression", 0.001, 0.781007, 0.703285 ])

x.add_row(["BOW", "LogisticRegression", 0.01, 0.86734, 0.712018 ])

x.add_row(["BOW", "LogisticRegression", 0.01, 0.96794, 0.66939 ])
```

```
x.add_row(["BOW", "LogisticRegression", 1, 0.99974, 0.63817 ])
x.add_row(["BOW", "LogisticRegression", 10, 1, 0.627842 ])
x.add_row(["BOW", "LogisticRegression", 100000, 0.99999, 0.625248 ])
print(x)
```

-		+		+	
_	 Vectorizer	Algorithm	Hyper parameter	Train AUC	Test AUC
-	BOW BOW BOW BOW (Best) BOW BOW BOW BOW	LogisticRegression LogisticRegression LogisticRegression LogisticRegression LogisticRegression LogisticRegression LogisticRegression LogisticRegression	1e-05 0.0001 0.001 0.01 0.1 1 10	0.57485 0.66466 0.781007 0.86734 0.96794 0.99974 1 0.99999	0.56908 0.639418 0.703285 0.712018 0.66939 0.63817 0.627842 0.625248
-					

When class_weight="Balance"

```
In [178]:
```

```
# Please compare all your models using Prettytable library
from prettytable import PrettyTable

x = PrettyTable()

x.field_names = ["Vectorizer", "Algorithm", "Hyper parameter", "Train AUC", "Test AUC"]

x.add_row(["BOW", "LogisticRegression", 0.00001, 0.576555, 0.585418 ])
x.add_row(["BOW", "LogisticRegression", 0.0001, 0.674211, 0.664531 ])
x.add_row(["BOW ", "LogisticRegression", 0.001, 0.777786, 0.716399 ])
x.add_row(["BOW", "LogisticRegression", 0.01, 0.839597, 0.726001 ])
x.add_row(["BOW", "LogisticRegression", 0.1, 0.943544, 0.677313 ])
x.add_row(["BOW", "LogisticRegression", 1, 0.995350, 0.644916 ])
x.add_row(["BOW", "LogisticRegression", 1, 0.999999, 0.628931 ])
print(x)
```

Vectorizer	Algorithm	Hyper parameter	Train AUC	Test AUC
BOW BOW BOW BOW BOW BOW BOW BOW BOW	LogisticRegression	1e-05	0.576555	0.585418
	LogisticRegression	0.0001	0.674211	0.664531
	LogisticRegression	0.001	0.777786	0.716399
	LogisticRegression	0.01	0.839597	0.726001
	LogisticRegression	0.1	0.943544	0.677313
	LogisticRegression	1	0.99535	0.644916
	LogisticRegression	10	0.99999	0.628931

In [139]:

```
# we are writing our own function for predict, with defined thresould
# we will pick a threshold that will give the least fpr

def predict(proba, threshould, fpr, tpr):

    t = threshould[np.argmax(tpr*(1-fpr))]

# (tpr*(1-fpr)) will be maximum if your fpr is very low and tpr is very high

print("the maximum value of tpr*(1-fpr)", max(tpr*(1-fpr)), "for threshold", np.round(t,3))
predictions = []

for i in proba:
    if i>=t:
        predictions.append(1)
    else:
        predictions.append(0)
return predictions
```

```
In [140]:
```

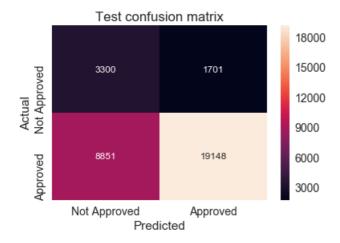
```
# https://scikit-learn.org/stable/modules/generated/sklearn.metrics.confusion_matrix.html
print("="*100)
from sklearn.metrics import confusion_matrix
print("Train confusion matrix")
print(confusion_matrix(y_train, predict(y_train_bow_pred, tr_thresholds, train_fpr, train_tpr)))
print("Test confusion matrix")
print(confusion_matrix(y_test, predict(y_test_bow_pred, te_thresholds, test_fpr, test_tpr)))

Train confusion matrix
the maximum value of tpr*(1-fpr) 0.5799920616124823 for threshold 0.824
[[ 7637 2516]
        [13014 43833]]
Test confusion matrix
the maximum value of tpr*(1-fpr) 0.4512715800352197 for threshold 0.848
[[ 3300 1701]
        [ 8851 19148]]
```

```
In [141]:
import seaborn as snTr
import seaborn as snTe
import pandas as pdH
import matplotlib.pyplot as pltTr
import matplotlib.pyplot as pltTe
# https://stackoverflow.com/questions/35572000/how-can-i-plot-a-confusion-matrix
arrayTr=confusion matrix(y train, predict(y train bow pred, tr thresholds, train fpr, train tpr))
df cmTr = pdH.DataFrame(arrayTr, range(2), range(2))
#print(arrayTr)
# https://stackoverflow.com/questions/32723798/how-do-i-add-a-title-to-seaborn-heatmap
axTr = pltTr.axes()
snTr.set(font scale=1.4)#for label size
# https://seaborn.pydata.org/generated/seaborn.heatmap.html
snTr.heatmap(df cmTr, annot=True,annot kws={"size": 12},fmt="d",ax=axTr)# font size, format in
digit
labels=['Not Approved','Approved']
axTr.set xticklabels(labels)
axTr.set yticklabels(labels)
#Suggestion 4.Label confusion matrix heatmap with actual and predicted labels.
pltTr.title("Train confusion matrix")
pltTr.xlabel("Predicted")
pltTr.ylabel("Actual")
pltTr.show()
# https://stackoverflow.com/questions/50947776/plot-two-seaborn-heatmap-graphs-side-by-side
#fig, ax =plt.subplots(1,1)
# https://stackoverflow.com/questions/35572000/how-can-i-plot-a-confusion-matrix
arrayTe=confusion matrix(y test, predict(y test bow pred, te thresholds, test fpr, test tpr))
df_cmTe = pdH.DataFrame(arrayTe, range(2), range(2))
axTe = pltTe.axes()
snTe.set(font_scale=1.4)#for label size
# https://seaborn.pydata.org/generated/seaborn.heatmap.html
snTe.heatmap(df cmTe, annot=True,annot kws={"size": 12},fmt="d",ax=axTe) # font size, format in
digit
#Suggestion 4.Label confusion matrix heatmap with actual and predicted labels.
axTe.set xticklabels(labels)
axTe.set_yticklabels(labels)
pltTe.title("Test confusion matrix")
pltTe.xlabel("Predicted")
pltTe.ylabel("Actual")
pltTe.show()
```



the maximum value of tpr*(1-fpr) 0.4512715800352197 for threshold 0.848



Task 2

- 1. [Task-1] Logistic Regression(either SGDClassifier with log loss, or LogisticRegression) on these feature sets
 - Set 2: categorical, numerical features + project_title(TFIDF)+ preprocessed_eassay (`TFIDF with bi-grams` with
 `min df=10` and `max features=5000`)

In [142]:

```
# merge two sparse matrices: https://stackoverflow.com/a/19710648/4084039
from scipy.sparse import hstack
X_tr_tfidf = hstack((X_train_text_tfidf, X_train_title_tfidf, X_train_state_ohe, X_train_clean_ohe
, X_train_cleanSub_ohe, X_train_grade_ohe, X_train_teacher_ohe, X_train_prjResSum_ohe,
X_train_quantity_norm, X_train_TprevPrj_norm, X_train_price_norm)).tocsr()
X_te_tfidf = hstack((X_test_text_tfidf, X_test_title_tfidf , X_test_state_ohe, X_test_clean_ohe, X_
test_cleanSub_ohe, X_test_grade_ohe, X_test_teacher_ohe, X_test_prjResSum_ohe,
X_test_quantity_norm, X_test_TprevPrj_norm, X_test_price_norm)).tocsr()
print("Final Data matrix | tfidf")
print(X_tr_tfidf.shape, y_train.shape)
print(X_te_tfidf.shape, y_test.shape)
print("="*100)
Final Data matrix | tfidf
(67000, 28703) (67000,)
(33000, 28703) (33000,)
                                                                                                ₩ ▶
```

```
#code source:
http://occam.olin.edu/sites/default/files/DataScienceMaterials/machine learning lecture 2/Machine%
rning%20Lecture%202.html
from sklearn.model selection import train test split
#from sklearn.grid_search import GridSearchCV
from sklearn.model selection import GridSearchCV
from sklearn.datasets import *
from sklearn.linear_model import LogisticRegression
c range=[10**-5, 10**-4, 10**-2, 10**0, 10**2, 10**4, 10**5]
param_grid=dict(C=c_range)
#Using GridSearchCV
# https://scikit-learn.org/stable/modules/generated/sklearn.model selection.GridSearchCV.html
model = GridSearchCV(LogisticRegression(class weight="balanced"), param grid, scoring = 'f1', cv=5)
model.fit(X tr tfidf, y train)
print(model.best estimator )
print(model.score(X te tfidf, y test))
4
LogisticRegression(C=100000, class weight='balanced', dual=False,
          fit_intercept=True, intercept_scaling=1, max_iter=100,
          multi class='ovr', n jobs=1, penalty='12', random state=None,
          solver='liblinear', tol=0.0001, verbose=0, warm start=False)
0.8433703847075522
In [144]:
train_tf_auc= model.cv_results_['mean_train_score']
train tf auc std= model.cv results ['std train score']
cv_tf_auc = model.cv_results_['mean_test_score']
cv_tf_auc_std= model.cv_results_['std_test_score']
CC = []
from math import log
CC = [np.log10(x) \text{ for } x \text{ in } c \text{ range}]
print (CC)
plt.plot(CC, train_tf_auc, label='Train AUC')
# this code is copied from here: https://stackoverflow.com/a/48803361/4084039
plt.gca().fill_between(CC,train_tf_auc - train_tf_auc_std,train_tf_auc + train_tf_auc_std,alpha=0.2
,color='darkblue')
plt.plot(CC, cv_tf_auc, label='CV AUC')
# this code is copied from here: https://stackoverflow.com/a/48803361/4084039
plt.gca().fill_between(CC,cv_tf_auc - cv_tf_auc_std,cv_tf_auc + cv_tf_auc_std,alpha=0.2,color='dark
orange')
plt.scatter(CC, train_tf_auc, label='Train AUC points')
```

[-5.0, -4.0, -2.0, 0.0, 2.0, 4.0, 5.0]

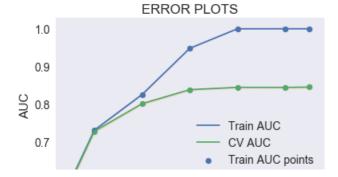
plt.xlabel("C: hyperparameter")

plt.legend()

plt.grid()
plt.show()

plt.ylabel("AUC")

plt.title("ERROR PLOTS")



plt.scatter(CC, cv_tf_auc, label='CV AUC points')

```
0.6 CV AUC points

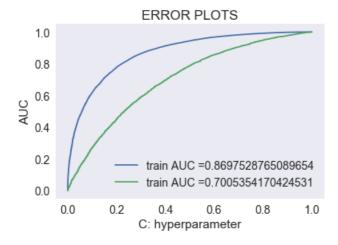
-4 -2 0 2 4
C: hyperparameter
```

In [145]:

```
best_tuned_parameters = [{'C': [0.1]}]
```

In [146]:

```
# https://scikit-
learn.org/stable/modules/generated/sklearn.metrics.roc curve.html#sklearn.metrics.roc curve
from sklearn.metrics import roc curve, auc
model = GridSearchCV(LogisticRegression(), best tuned parameters)
model.fit(X_tr_tfidf, y_train)
# roc_auc_score(y_true, y_score) the 2nd parameter should be probability estimates of the positive
class
# not the predicted outputs
print(model.best estimator )
print(model.score(X_te_tfidf, y_test))
y train tf pred = model.predict proba(X tr tfidf)[:,1]
y_test_tf_pred = model.predict_proba(X_te_tfidf)[:,1]
train_fpr, train_tpr, tr_thresholds = roc_curve(y_train, y_train_tf_pred)
test_fpr, test_tpr, te_thresholds = roc_curve(y_test, y_test_tf_pred)
plt.plot(train_fpr, train_tpr, label="train AUC ="+str(auc(train_fpr, train_tpr)))
plt.plot(test fpr, test tpr, label="train AUC ="+str(auc(test fpr, test tpr)))
plt.legend()
plt.xlabel("C: hyperparameter")
plt.ylabel("AUC")
plt.title("ERROR PLOTS")
plt.arid()
plt.show()
```



In [147]:

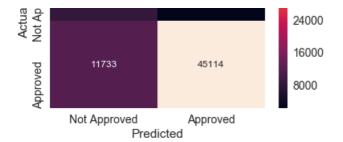
```
# https://scikit-learn.org/stable/modules/generated/sklearn.metrics.confusion_matrix.html
print("="*100)
from sklearn.metrics import confusion_matrix
print("Train confusion matrix")
print(confusion_matrix(y_train, predict(y_train_tf_pred, tr_thresholds, train_fpr, train_tpr)))
print("Test_confusion_matrix")
print(confusion_matrix(y_test, predict(y_test_tf_pred, te_thresholds, test_fpr, test_tpr)))
```

In [148]:

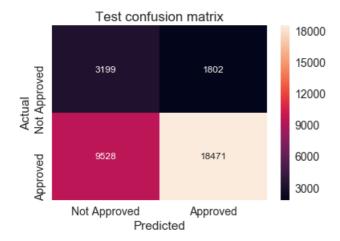
```
import seaborn as snTr
import seaborn as snTe
import pandas as pdH
import matplotlib.pyplot as pltTr
import matplotlib.pyplot as pltTe
# https://stackoverflow.com/questions/35572000/how-can-i-plot-a-confusion-matrix
arrayTr=confusion_matrix(y_train, predict(y_train_tf_pred, tr_thresholds, train_fpr, train_tpr))
df cmTr = pdH.DataFrame(arrayTr,range(2),range(2))
#print(arrayTr)
# https://stackoverflow.com/questions/32723798/how-do-i-add-a-title-to-seaborn-heatmap
axTr = pltTr.axes()
snTr.set(font_scale=1.4)#for label size
# https://seaborn.pydata.org/generated/seaborn.heatmap.html
snTr.heatmap(df cmTr, annot=True,annot kws={"size": 12},fmt="d",ax=axTr) # font size, format in
digit
labels=['Not Approved','Approved']
axTr.set xticklabels(labels)
axTr.set_yticklabels(labels)
#Suggestion 4.Label confusion matrix heatmap with actual and predicted labels.
pltTr.title("Train confusion matrix")
pltTr.xlabel("Predicted")
pltTr.ylabel("Actual")
pltTr.show()
# https://stackoverflow.com/questions/50947776/plot-two-seaborn-heatmap-graphs-side-by-side
#fig, ax =plt.subplots(1,1)
# https://stackoverflow.com/questions/35572000/how-can-i-plot-a-confusion-matrix
arrayTe=confusion_matrix(y_test, predict(y_test_tf_pred, te_thresholds, test_fpr, test tpr))
df cmTe = pdH.DataFrame(arrayTe, range(2), range(2))
axTe = pltTe.axes()
snTe.set(font scale=1.4)#for label size
# https://seaborn.pydata.org/generated/seaborn.heatmap.html
snTe.heatmap(df cmTe, annot=True,annot kws={"size": 12},fmt="d",ax=axTe)# font size, format in
#Suggestion 4. Label confusion matrix heatmap with actual and predicted labels.
axTe.set xticklabels(labels)
axTe.set yticklabels(labels)
pltTe.title("Test confusion matrix")
pltTe.xlabel("Predicted")
pltTe.ylabel("Actual")
pltTe.show()
```

the maximum value of tpr*(1-fpr) 0.6232053351727436 for threshold 0.825





the maximum value of tpr*(1-fpr) 0.4219930255886035 for threshold 0.855



Task 3

- 1. [Task-1] Logistic Regression(either SGDClassifier with log loss, or LogisticRegression) on these feature sets
 - Set 3: categorical, numerical features + project title(AVG W2V)+ preprocessed eassay (AVG W2V)

In [149]:

here

In [152]:

```
#code source:
http://occam.olin.edu/sites/default/files/DataScienceMaterials/machine_learning_lecture_2/Machine%2
rning%20Lecture%202.html
from sklearn.model_selection import train_test_split
#from sklearn.grid_search import GridSearchCV
from sklearn.model_selection import GridSearchCV
```

```
from sklearn.datasets import *
from sklearn.linear_model import LogisticRegression

c_range=[10**-5, 10**-4, 10**-2, 10**0, 10**2, 10**4, 10**5]
param_grid=dict(C=c_range)

#Using GridSearchCV
# https://scikit-learn.org/stable/modules/generated/sklearn.model_selection.GridSearchCV.html
model = GridSearchCV(LogisticRegression(class_weight="balanced"), param_grid, scoring = 'f1', cv=5)
model.fit(X_tr_avgW2V, y_train)

print(model.best_estimator_)
print(model.score(X_te_avgW2V, y_test))

#Using GridSearchCV
# https://scikit-learn.org/stable/modules/generated/sklearn.model_selection.GridSearchCV.html
model = GridSearchCV(LogisticRegression(class_weight="balanced"), param_grid, scoring = 'f1', cv=5)
model.fit(X_tr_avgW2V, y_train)

print(model.best_estimator_)
print(model.score(X_te_avgW2V, y_test))

#Using GridSearchCV
# https://scikit-learn.org/stable/modules/generated/sklearn.model_selection.GridSearchCV.html
model = GridSearchCV (LogisticRegression(class_weight="balanced"), param_grid, scoring = 'f1', cv=5)
model.fit(X_tr_avgW2V, y_train)

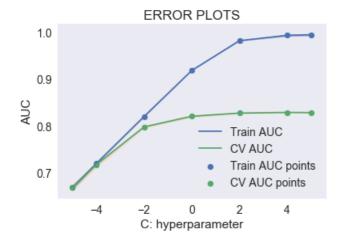
print(model.best_estimator_)
print(model.score(X_te_avgW2V, y_test))

#Using GridSearchCV
# https://scikit-learn.org/stable/modules/generated/sklearn.model_selection.GridSearchCV.html
model = GridSearchCV.html
model = GridSearchCV
# https://scikit-learn.org/stable/modules/generated/sklearn.model_selection.GridSearchCV.html
model = GridSearchCV
# https://scikit-learn.org/stable/modules/generated/sklearn.model
```

In [153]:

```
train avgW2V auc= model.cv results ['mean train score']
train avgW2V auc std= model.cv results ['std train score']
cv_avgW2V_auc = model.cv_results_['mean_test_score']
cv avgW2V auc std= model.cv results ['std test score']
CC = []
from math import log
CC = [np.log10(x) \text{ for } x \text{ in } c \text{ range}]
print(CC)
plt.plot(CC, train_avgW2V_auc, label='Train AUC')
# this code is copied from here: https://stackoverflow.com/a/48803361/4084039
plt.gca().fill between(CC,train avgW2V auc - train avgW2V auc std,train avgW2V auc + train avgW2V a
uc_std,alpha=0.2,color='darkblue')
plt.plot(CC, cv_avgW2V_auc, label='CV AUC')
# this code is copied from here: https://stackoverflow.com/a/48803361/4084039
plt.gca().fill between (CC,cv avgW2V auc - cv avgW2V auc std,cv avgW2V auc +
cv avgW2V auc std,alpha=0.2,color='darkorange')
plt.scatter(CC, train avgW2V auc, label='Train AUC points')
plt.scatter(CC, cv_avgW2V_auc, label='CV AUC points')
plt.legend()
plt.xlabel("C: hyperparameter")
plt.ylabel("AUC")
plt.title("ERROR PLOTS")
plt.grid()
plt.show()
```

[-5.0, -4.0, -2.0, 0.0, 2.0, 4.0, 5.0]

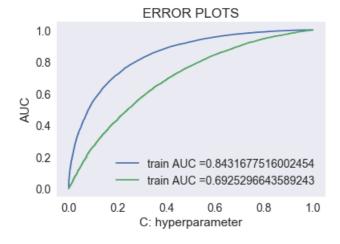


```
In [154]:
```

```
best_tuned_parameters = [{'C': [0.1]}]
```

In [155]:

```
# https://scikit-
learn.org/stable/modules/generated/sklearn.metrics.roc curve.html#sklearn.metrics.roc curve
from sklearn.metrics import roc curve, auc
model = GridSearchCV(LogisticRegression(), best tuned parameters)
model.fit(X tr avgW2V, y train)
# roc auc score(y true, y score) the 2nd parameter should be probability estimates of the positive
class
# not the predicted outputs
print(model.best estimator )
print(model.score(X_te_avgW2V, y_test))
y train avgW2V pred = model.predict proba(X tr avgW2V)[:,1]
y_test_avgW2V_pred = model.predict_proba(X_te_avgW2V)[:,1]
train fpr, train tpr, tr thresholds = roc curve(y train, y train avgW2V pred)
test fpr, test_tpr, te_thresholds = roc_curve(y_test, y_test_avgW2V_pred)
plt.plot(train fpr, train tpr, label="train AUC ="+str(auc(train fpr, train tpr)))
plt.plot(test fpr, test tpr, label="train AUC ="+str(auc(test fpr, test tpr)))
plt.legend()
plt.xlabel("C: hyperparameter")
plt.ylabel("AUC")
plt.title("ERROR PLOTS")
plt.grid()
plt.show()
```



In [157]:

```
# https://scikit-learn.org/stable/modules/generated/sklearn.metrics.confusion_matrix.html
print("="*100)
from sklearn.metrics import confusion_matrix
print("Train confusion matrix")
print(confusion_matrix(y_train, predict(y_train_avgW2V_pred, tr_thresholds, train_fpr,
train_tpr)))
print("Test confusion matrix")
print(confusion_matrix(y_test, predict(y_test_avgW2V_pred, te_thresholds, test_fpr, test_tpr)))
```

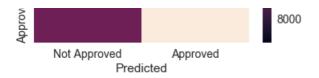
Train confusion matrix the maximum value of tpr*(1-fpr) 0.584942696479297 for threshold 0.835 [[7829 2324]

```
In [158]:
```

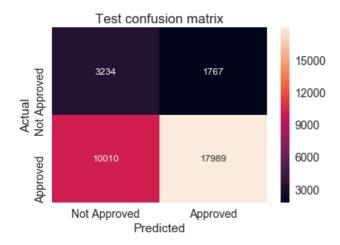
```
import seaborn as snTr
import seaborn as snTe
import pandas as pdH
import matplotlib.pyplot as pltTr
import matplotlib.pyplot as pltTe
# https://stackoverflow.com/questions/35572000/how-can-i-plot-a-confusion-matrix
arrayTr=confusion_matrix(y_train, predict(y_train_avgW2V_pred, tr_thresholds, train_fpr, train_tpr
df cmTr = pdH.DataFrame(arrayTr, range(2), range(2))
#print(arrayTr)
# https://stackoverflow.com/questions/32723798/how-do-i-add-a-title-to-seaborn-heatmap
axTr = pltTr.axes()
snTr.set(font scale=1.4)#for label size
# https://seaborn.pydata.org/generated/seaborn.heatmap.html
snTr.heatmap(df cmTr, annot=True,annot kws={"size": 12},fmt="d",ax=axTr)# font size, format in
digit
labels=['Not Approved','Approved']
axTr.set xticklabels(labels)
axTr.set_yticklabels(labels)
#Suggestion 4.Label confusion matrix heatmap with actual and predicted labels.
pltTr.title("Train confusion matrix")
pltTr.xlabel("Predicted")
pltTr.ylabel("Actual")
pltTr.show()
# https://stackoverflow.com/questions/50947776/plot-two-seaborn-heatmap-graphs-side-by-side
#fig, ax =plt.subplots(1,1)
# https://stackoverflow.com/questions/35572000/how-can-i-plot-a-confusion-matrix
arrayTe=confusion matrix(y test, predict(y test avgW2V pred, te thresholds, test fpr, test tpr))
df cmTe = pdH.DataFrame(arrayTe, range(2), range(2))
axTe = pltTe.axes()
snTe.set(font scale=1.4)#for label size
# https://seaborn.pydata.org/generated/seaborn.heatmap.html
snTe.heatmap(df cmTe, annot=True,annot kws={"size": 12},fmt="d",ax=axTe) # font size, format in
diait
#Suggestion 4.Label confusion matrix heatmap with actual and predicted labels.
axTe.set xticklabels(labels)
axTe.set yticklabels(labels)
pltTe.title("Test confusion matrix")
pltTe.xlabel("Predicted")
pltTe.ylabel("Actual")
pltTe.show()
```

the maximum value of tpr*(1-fpr) 0.584942696479297 for threshold 0.835





the maximum value of tpr*(1-fpr) 0.4154776459258668 for threshold 0.858



Task 4

- 1. [Task-1] Logistic Regression(either SGDClassifier with log loss, or LogisticRegression) on these feature sets
 - Set 4: categorical, numerical features + project_title(TFIDF W2V)+ preprocessed_essay (TFIDF W2V)

In [159]:

In [160]:

```
#code source:
http://occam.olin.edu/sites/default/files/DataScienceMaterials/machine_learning_lecture_2/Machine%i
rning%20Lecture%202.html
from sklearn.model_selection import train_test_split
#from sklearn.grid_search import GridSearchCV
from sklearn.model_selection import GridSearchCV
from sklearn.datasets import *
from sklearn.linear_model import LogisticRegression

c_range=[10**-5, 10**-4, 10**-2, 10**0, 10**2, 10**4, 10**5]
param_grid=dict(C=c_range)

#Using GridSearchCV
# https://scikit-learn.org/stable/modules/generated/sklearn.model_selection.GridSearchCV.html
model = GridSearchCV(LogisticRegression(class weight="balanced"), param grid, scoring = 'f1', cv=5)
```

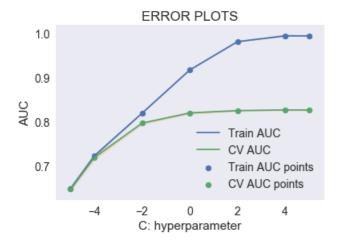
```
model.fit(X_tr_tfidf_W2V, y_train)

print(model.best_estimator_)
print(model.score(X_te_tfidf_W2V, y_test))
```

In [161]:

```
train tfidf w2v auc= model.cv results ['mean train score']
train tfidf w2v auc std= model.cv results ['std train score']
cv tfidf w2v auc = model.cv results ['mean test score']
cv tfidf w2v auc std= model.cv results ['std test score']
CC = []
from math import log
CC = [np.log10(x) for x in c_range]
print(CC)
plt.plot(CC, train_tfidf_w2v_auc, label='Train AUC')
# this code is copied from here: https://stackoverflow.com/a/48803361/4084039
\verb|plt.gca().fill between(CC, train\_tfidf\_w2v\_auc - train\_tfidf\_w2v\_auc\_std, train\_tfidf\_w2v\_auc + train\_tfidf\_w2v\_auc + train\_tfidf\_w2v\_auc + train\_tfidf\_w2v\_auc\_std, train\_tfidf\_w2v\_auc + train\_tfidf\_w2v\_auc\_std, train\_tfidf\_w2v\_auc + train\_tfidf\_w2v\_auc\_std, train\_tf
tfidf w2v auc std,alpha=0.2,color='darkblue')
plt.plot(CC, cv_tfidf_w2v_auc, label='CV AUC')
# this code is copied from here: https://stackoverflow.com/a/48803361/4084039
plt.gca().fill_between(CC,cv_tfidf_w2v_auc - cv_tfidf_w2v_auc_std,cv_tfidf_w2v_auc + cv_tfidf_w2v_a
uc std,alpha=0.2,color='darkorange')
plt.scatter(CC, train_tfidf_w2v_auc, label='Train AUC points')
plt.scatter(CC, cv tfidf w2v auc, label='CV AUC points')
plt.legend()
plt.xlabel("C: hyperparameter")
plt.ylabel("AUC")
plt.title("ERROR PLOTS")
plt.grid()
plt.show()
```

[-5.0, -4.0, -2.0, 0.0, 2.0, 4.0, 5.0]



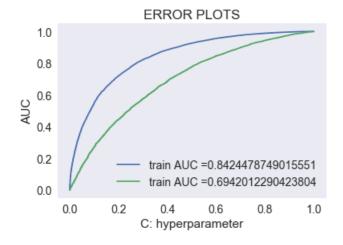
In [162]:

```
best_tuned_parameters = [{'C': [0.1]}]
```

In [164]:

```
# https://scikit-
learn.org/stable/modules/generated/sklearn.metrics.roc_curve.html#sklearn.metrics.roc_curve
```

```
from sklearn.metrics import roc_curve, auc
model = GridSearchCV(LogisticRegression(), best tuned parameters)
model.fit(X_tr_tfidf_W2V, y_train)
# roc auc score(y true, y score) the 2nd parameter should be probability estimates of the positive
class
# not the predicted outputs
print(model.best_estimator_)
print(model.score(X te tfidf W2V, y test))
y train tfidf w2v pred = model.predict proba(X tr tfidf W2V)[:,1]
y test tfidf w2v pred = model.predict proba(X te tfidf W2V)[:,1]
train_fpr, train_tpr, tr_thresholds = roc_curve(y_train, y_train_tfidf_w2v_pred)
test fpr, test tpr, te thresholds = roc curve(y test, y test tfidf w2v pred)
plt.plot(train_fpr, train_tpr, label="train AUC ="+str(auc(train fpr, train tpr)))
plt.plot(test fpr, test tpr, label="train AUC ="+str(auc(test fpr, test tpr)))
plt.legend()
plt.xlabel("C: hyperparameter")
plt.ylabel("AUC")
plt.title("ERROR PLOTS")
plt.grid()
plt.show()
```

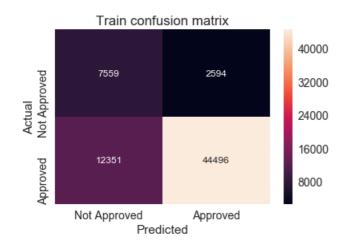


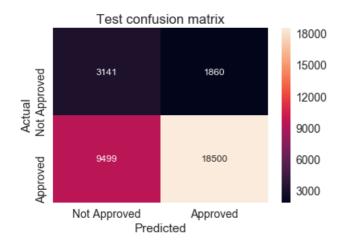
In [166]:

```
# https://scikit-learn.org/stable/modules/generated/sklearn.metrics.confusion_matrix.html
print("="*100)
from sklearn.metrics import confusion_matrix
print("Train confusion matrix")
print(confusion_matrix(y_train, predict(y_train_tfidf_w2v_pred, tr_thresholds, train_fpr,
train_tpr)))
print("Test confusion matrix")
print(confusion_matrix(y_test, predict(y_test_tfidf_w2v_pred, te_thresholds, test_fpr, test_tpr)))
```

```
import seaborn as snTr
import seaborn as snTe
import pandas as pdH
import matplotlib.pyplot as pltTr
import matplotlib.pyplot as pltTe
# https://stackoverflow.com/questions/35572000/how-can-i-plot-a-confusion-matrix
arrayTr=confusion matrix(y train, predict(y train tfidf w2v pred, tr thresholds, train fpr,
train tpr))
df cmTr = pdH.DataFrame(arrayTr, range(2), range(2))
#print(arrayTr)
# https://stackoverflow.com/questions/32723798/how-do-i-add-a-title-to-seaborn-heatmap
axTr = pltTr.axes()
snTr.set(font_scale=1.4)#for label size
# https://seaborn.pydata.org/generated/seaborn.heatmap.html
snTr.heatmap(df cmTr, annot=True,annot kws={"size": 12},fmt="d",ax=axTr)# font size, format in
digit
labels=['Not Approved','Approved']
axTr.set xticklabels(labels)
axTr.set yticklabels(labels)
#Suggestion 4.Label confusion matrix heatmap with actual and predicted labels.
pltTr.title("Train confusion matrix")
pltTr.xlabel("Predicted")
pltTr.ylabel("Actual")
pltTr.show()
# https://stackoverflow.com/questions/50947776/plot-two-seaborn-heatmap-graphs-side-by-side
#fig, ax =plt.subplots(1,1)
# https://stackoverflow.com/questions/35572000/how-can-i-plot-a-confusion-matrix
arrayTe=confusion matrix(y test, predict(y test tfidf w2v pred, te thresholds, test fpr, test tpr))
df cmTe = pdH.DataFrame(arrayTe, range(2), range(2))
axTe = pltTe.axes()
snTe.set(font scale=1.4)#for label size
# https://seaborn.pydata.org/generated/seaborn.heatmap.html
snTe.heatmap(df cmTe, annot=True,annot kws={"size": 12},fmt="d",ax=axTe)# font size, format in
digit
#Suggestion 4.Label confusion matrix heatmap with actual and predicted labels.
axTe.set_xticklabels(labels)
axTe.set_yticklabels(labels)
pltTe.title("Test confusion matrix")
pltTe.xlabel("Predicted")
pltTe.ylabel("Actual")
pltTe.show()
```

the maximum value of tpr*(1-fpr) 0.5827514733064767 for threshold 0.823





2.5 Logistic Regression with added Features 'Set 5'

```
In [ ]:
```

```
# please write all the code with proper documentation, and proper titles for each subsection
# go through documentations and blogs before you start coding
# first figure out what to do, and then think about how to do.
# reading and understanding error messages will be very much helpfull in debugging your code
# when you plot any graph make sure you use
   # a. Title, that describes your plot, this will be very helpful to the reader
    # b. Legends if needed
    # c. X-axis label
    # d. Y-axis label
```

- 1. [Task-2] Apply Logistic Regression on the below feature set Set 5 by finding the best hyper parameter as suggested in step 2 and step 3.
- 2. Consider these set of features Set 5:
 - · school_state : categorical data
 - clean_categories : categorical data
 - clean_subcategories : categorical data
 - · project_grade_category :categorical data
 - teacher prefix : categorical data
 - · quantity: numerical data
 - teacher_number_of_previously_posted_projects : numerical data
 - price: numerical data
 - sentiment score's of each of the essay : numerical data
 - · number of words in the title: numerical data
 - number of words in the combine essays: numerical data

And apply the Logistic regression on these features by finding the best hyper paramter as suggested in step 2 and step 3

In [168]:

```
# merge two sparse matrices: https://stackoverflow.com/a/19710648/4084039
from scipy.sparse import hstack
X_tr_SET5 = hstack((X_train_state_ohe, X_train_clean_ohe, X_train_cleanSub_ohe, X_train_grade_ohe,
X_train_teacher_ohe, X_train_quantity_norm, X_train_TprevPrj_norm,
X train price norm, X train neg norm, X train pos norm, X train neu norm, X train compound norm, X train
title wc norm, X train essay wc norm)).tocsr()
X te SET5 = hstack((X test state ohe, X test clean ohe, X test cleanSub ohe, X test grade ohe, X te
st_teacher_ohe, X_test_quantity_norm, X_test_TprevPrj_norm, X_test_price_norm, X_test_neg_norm,
X test pos norm, X test neu norm, X test compound norm, X test title wc norm, X test essay wc norm
)).tocsr()
print("Final Data matrix | SET 5")
print(X_tr_SET5.shape, y_train.shape)
print(X_te_SET5.shape, y_test.shape)
print("="*100)
                                                                                                 •
```

```
Final Data matrix | SET 5
(67000, 108) (67000,)
```

In [169]:

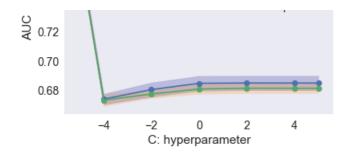
```
#code source:
http://occam.olin.edu/sites/default/files/DataScienceMaterials/machine learning lecture 2/Machine%2
rning%20Lecture%202.html
from sklearn.model_selection import train test split
#from sklearn.grid search import GridSearchCV
from sklearn.model_selection import GridSearchCV
from sklearn.datasets import
from sklearn.linear model import LogisticRegression
c range=[10**-5, 10**-4, 10**-2, 10**0, 10**2, 10**4, 10**5]
param grid=dict(C=c range)
#Using GridSearchCV
# https://scikit-learn.org/stable/modules/generated/sklearn.model_selection.GridSearchCV.html
model = GridSearchCV(LogisticRegression(class weight="balanced"), param grid, scoring = 'f1', cv=5)
model.fit(X_tr_SET5, y_train)
print(model.best estimator )
print (model.score (X te SET5, y test))
4
```

In [170]:

```
train SET5 auc= model.cv results ['mean train score']
train SET5 auc std= model.cv results ['std train score']
cv SET5 auc = model.cv results ['mean test score']
cv SET5 auc std= model.cv results ['std test score']
CC = []
from math import log
CC = [np.log10(x)  for x  in c  range]
print(CC)
plt.plot(CC, train SET5 auc, label='Train AUC')
# this code is copied from here: https://stackoverflow.com/a/48803361/4084039
plt.gca().fill between(CC,train SET5 auc - train SET5 auc std,train SET5 auc + train SET5 auc std,a
lpha=0.2,color='darkblue')
plt.plot(CC, cv SET5 auc, label='CV AUC')
# this code is copied from here: https://stackoverflow.com/a/48803361/4084039
plt.gca().fill_between(CC,cv_SET5_auc - cv_SET5_auc_std,cv_SET5_auc + cv_SET5_auc_std,alpha=0.2,col
or='darkorange')
plt.scatter(CC, train_SET5_auc, label='Train AUC points')
plt.scatter(CC, cv SET5 auc, label='CV AUC points')
plt.legend()
plt.xlabel("C: hyperparameter")
plt.ylabel("AUC")
plt.title("ERROR PLOTS")
plt.grid()
plt.show()
```

[-5.0, -4.0, -2.0, 0.0, 2.0, 4.0, 5.0]



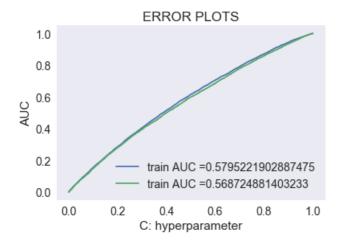


In [171]:

```
best_tuned_parameters = [{'C': [0.1]}]
```

In [172]:

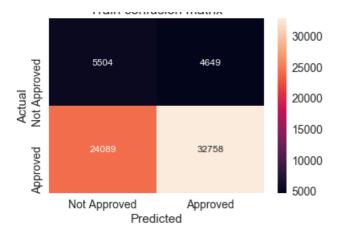
```
# https://scikit-
learn.org/stable/modules/generated/sklearn.metrics.roc curve.html#sklearn.metrics.roc curve
from sklearn.metrics import roc curve, auc
model = GridSearchCV(LogisticRegression(), best tuned parameters)
model.fit(X tr SET5, y train)
# roc auc score(y true, y score) the 2nd parameter should be probability estimates of the positive
class
# not the predicted outputs
print(model.best estimator )
print(model.score(X te SET5, y test))
y train SET5 pred = model.predict proba(X tr SET5)[:,1]
y test SET5 pred = model.predict proba(X te SET5)[:,1]
train_fpr, train_tpr, tr_thresholds = roc_curve(y_train, y_train_SET5_pred)
test fpr, test tpr, te thresholds = roc curve(y test, y test SET5 pred)
plt.plot(train fpr, train tpr, label="train AUC ="+str(auc(train fpr, train tpr)))
plt.plot(test fpr, test tpr, label="train AUC ="+str(auc(test fpr, test tpr)))
plt.legend()
plt.xlabel("C: hyperparameter")
plt.ylabel("AUC")
plt.title("ERROR PLOTS")
plt.grid()
plt.show()
```



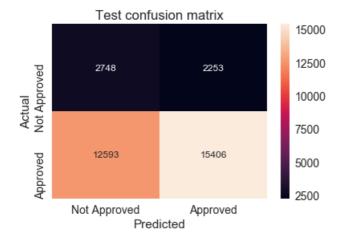
In [174]:

```
from sklearn.metrics import confusion matrix
print("Train confusion matrix")
print(confusion_matrix(y_train, predict(y_train_SET5_pred, tr_thresholds, train_fpr, train_tpr)))
print("Test confusion matrix")
print(confusion_matrix(y_test, predict(y_test_SET5_pred, te_thresholds, test_fpr, test_tpr)))
______
Train confusion matrix
the maximum value of tpr*(1-fpr) 0.31238765795496654 for threshold 0.847
[[ 5504 4649]
[24089 32758]]
Test confusion matrix
the maximum value of tpr*(1-fpr) 0.302348102114282 for threshold 0.849
[[ 2748 2253]
 [12593 15406]]
In [175]:
import seaborn as snTr
import seaborn as snTe
import pandas as pdH
import matplotlib.pyplot as pltTr
import matplotlib.pyplot as pltTe
# https://stackoverflow.com/questions/35572000/how-can-i-plot-a-confusion-matrix
arrayTr=confusion matrix(y train, predict(y train SET5 pred, tr thresholds, train fpr, train tpr))
df cmTr = pdH.DataFrame(arrayTr, range(2), range(2))
#print(arrayTr)
# https://stackoverflow.com/questions/32723798/how-do-i-add-a-title-to-seaborn-heatmap
axTr = pltTr.axes()
snTr.set(font scale=1.4)#for label size
# https://seaborn.pydata.org/generated/seaborn.heatmap.html
snTr.heatmap(df_cmTr, annot=True,annot_kws={"size": 12},fmt="d",ax=axTr)# font size, format in
digit
labels=['Not Approved','Approved']
axTr.set xticklabels(labels)
axTr.set yticklabels(labels)
#Suggestion 4.Label confusion matrix heatmap with actual and predicted labels.
pltTr.title("Train confusion matrix")
pltTr.xlabel("Predicted")
pltTr.ylabel("Actual")
pltTr.show()
# https://stackoverflow.com/questions/50947776/plot-two-seaborn-heatmap-graphs-side-by-side
#fig, ax =plt.subplots(1,1)
# https://stackoverflow.com/questions/35572000/how-can-i-plot-a-confusion-matrix
arrayTe=confusion_matrix(y_test, predict(y_test_SET5_pred, te_thresholds, test_fpr, test_tpr))
df_cmTe = pdH.DataFrame(arrayTe, range(2), range(2))
axTe = pltTe.axes()
snTe.set(font scale=1.4)#for label size
# https://seaborn.pydata.org/generated/seaborn.heatmap.html
snTe.heatmap(df cmTe, annot=True,annot kws={"size": 12},fmt="d",ax=axTe)# font size, format in
digit
#Suggestion 4.Label confusion matrix heatmap with actual and predicted labels.
axTe.set xticklabels(labels)
axTe.set yticklabels(labels)
pltTe.title("Test confusion matrix")
pltTe.xlabel("Predicted")
pltTe.ylabel("Actual")
pltTe.show()
```

the maximum value of tpr*(1-fpr) 0.31238765795496654 for threshold 0.847



the maximum value of tpr*(1-fpr) 0.302348102114282 for threshold 0.849



3. Conclusion

```
In [180]:
```

```
# Please compare all your models using Prettytable library
from prettytable import PrettyTable

x = PrettyTable()

x.field_names = ["Vectorizer", "Algorithm", "Hyper parameter", "Test AUC"]

x.add_row(["BOW", "Logistic Regression", 0.01, 0.726001 ])
x.add_row(["TFIDF", "Logistic Regression", 0.1, 0.700535 ])
x.add_row(["AVG W2V", "Logistic Regression", 0.1, 0.69252 ])
x.add_row(["TFIDF W2V", "Logistic Regression", 0.1, 0.694201 ])
x.add_row(["SET 5", "Logistic Regression", 0.1, 0.568724 ])

print(x)
```

+	Vectorizer	Algorithm	Hyper parameter	Test AUC
+ 	BOW TFIDF AVG W2V TFIDF W2V	Logistic Regression Logistic Regression Logistic Regression Logistic Regression	0.01 0.1 0.1 0.1	0.726001 0.700535 0.69252 0.694201
+	SET 5 	Logistic Regression +	0.1 +	0.568724

4. Summary

Step Tollowed

- · Preprocessing of Project subject categories Project subject subcategories project grade category teacher prefix Project essay Project title project resource summary
- · Numeric feature for Text no of words in essay length of each cell in essay no of words in Title length of each cell in Title no of words in Project resource summary lenght of each cell in Project resource summary
- Using Pretrained Models: Avg W2V
- · Computing Sentiment Scores for Project essay. Added below columns neg pos neu compound
- · Added all the features to project data
- Took data points for doing the assignment and separate the Class lable (Project_is_approved)
- · Splitting Data into Train (further split into Train and Cross validation) and Test.
- Making datamodel ready

text

- encoding of school_state is splited into Train,CV and Test vector
- encoding of clean category is splited into Train, CV and Test vector
- encoding of clean subcategory is splited into Train,CV and Test vector
- encoding of project grade category is splited into Train, CV and Test vect

or

- encoding of teacher prefix is splited into Train, CV and Test vector
- encoding of project resource summary is splited into Train, CV and Test ve

ctor

numeric

- encoding of quantity is splited into Train, CV and Test vector
- encoding of teacher number of previously posted projects is splited into Train, CV and Test vector
 - encoding of price is splited into Train, CV and Test vector
- encoding of sentimental score | neg, is splited into Train,CV and Test ve ctor
- encoding of sentimental score | pos, is splited into Train,CV and Test ve

ctor

- encoding of sentimental score | neu, is splited into Train,CV and Test ve

ctor

- encoding of sentimental score | compound, is splited into Train,CV and Te st vector

- encoding of title_words is splited into Train,CV and Test vector
- encoding of essay words is splited into Train, CV and Test vector
- encoding of project essay(BOW) is splited into Train,CV and Test vector
- encoding of project_title(BOW) is splited into Train,CV and Test vector
- encoding of project essay(TFIDF) is splited into Train, CV and Test vector
- encoding of project title(TFIDF) is splited into Train,CV and Test vector
- encoding of project essay(AVG W2V) is splited into Train,CV and Test vect or

or

- encoding of project title (AVG W2V) is splited into Train, CV and Test vect

- encoding of project_essay(TFIDF W2V) is splited into Train,CV and Test ve

ctor

- encoding of project title(TFIDF W2V) is splited into Train,CV and Test ve

•

ctor

4

For SET 1

Merging all the above features for SET 1

- Horizontally merging(with hstack) all categorical, numerical features + project_title(BOW) + preprocessed_essay (BOW)
- Fit a model on on train (on above merge features) data by using GridSearchCV(Logistic regression)
- . Draw a graph in Train and CV for varies values of C
- choose various value of Best_C and seeing the Best Test_AUC
- Since wanted to see difference between class_weight="Auto" and class_weight="Balance", made two pretty tableswith various values of C
- · Create Confusion matrix, in heatmap.

For SET 2

Merging all the above features for SET 2

- Horizontally merging(with hstack) all categorical, numerical features + project_title(TFIDF) + preprocessed_essay (TFIDF)
- Fit a model on on train (on above merge features) data by using GridSearchCV(Logistic regression)
- . Draw a graph in Train and CV for varies values of C
- . choose various value of Best_C and seeing the Best Test_AUC
- Create Confusion matrix, in heatmap.

For SET 3

Merging all the above features for SET 3

- Horizontally merging(with hstack) all categorical, numerical features + project_title(AVG W2V) + preprocessed_essay (AVG W2V)
- Fit a model on on train (on above merge features) data by using GridSearchCV(Logistic regression)
- Draw a graph in Train and CV for varies values of C
- choose Best_C from bestestimator function
- · Create Confusion matrix, in heatmap.

For SET 4

Merging all the above features for SET 4

- Horizontally merging(with hstack) all categorical, numerical features + project_title(TFIDF W2V) + preprocessed_essay (TFIDF W2V)
- Fit a model on on train (on above merge features) data by using GridSearchCV(Logistic regression)
- Draw a graph in Train and CV for varies values of C
- choose Best_C from bestestimator function
- · Create Confusion matrix, in heatmap.

For SET 5

Merging all the above features for SET 5

- Horizontally merging(with hstack) all categorical, numerical features + four sentiment score + word count for project title + word count for combine essay
- Fit a model on on train (on above merge features) data by using GridSearchCV(Logistic regression)
- Draw a graph in Train and CV for varies values of C
- choose Best_C from bestestimator function
- · Create Confusion matrix, in heatmap.